

Network Systems  
Science & Advanced  
Computing  
Biocomplexity Institute  
& Initiative  
University of Virginia

# Foresight and Analysis of Infectious Disease Threats to Virginia's Public Health

March 2<sup>nd</sup>, 2023

(data current to February 23<sup>rd</sup> – March 1<sup>st</sup>)

Biocomplexity Institute Technical report: TR BI-2023-22



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**BIOCOMPLEXITY** INSTITUTE

[biocomplexity.virginia.edu](https://biocomplexity.virginia.edu)

# About Us

- Biocomplexity Institute at the University of Virginia
  - Using big data and simulations to understand massively interactive systems and solve societal problems
- Over 20 years of crafting and analyzing infectious disease models
  - Pandemic response for Influenza, Ebola, Zika, and others



## Points of Contact

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# Overview

- **Goal:** Understand impact of current and emerging Infectious Disease threats to the Commonwealth of Virginia using modeling and analytics
- **Approach:**
  - Provide analyses and summaries of current infectious disease threats
  - Survey existing forecasts and trends in these threats
  - Analyze and summarize the current situation and trends of these threats in the broader context of the US and world.
  - Provide broader overview of other emerging threats

# Key Takeaways

Projecting future cases precisely is impossible and unnecessary.

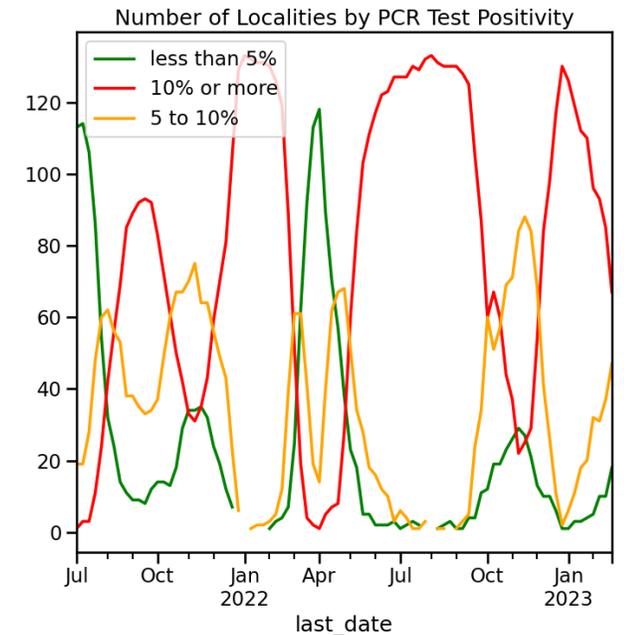
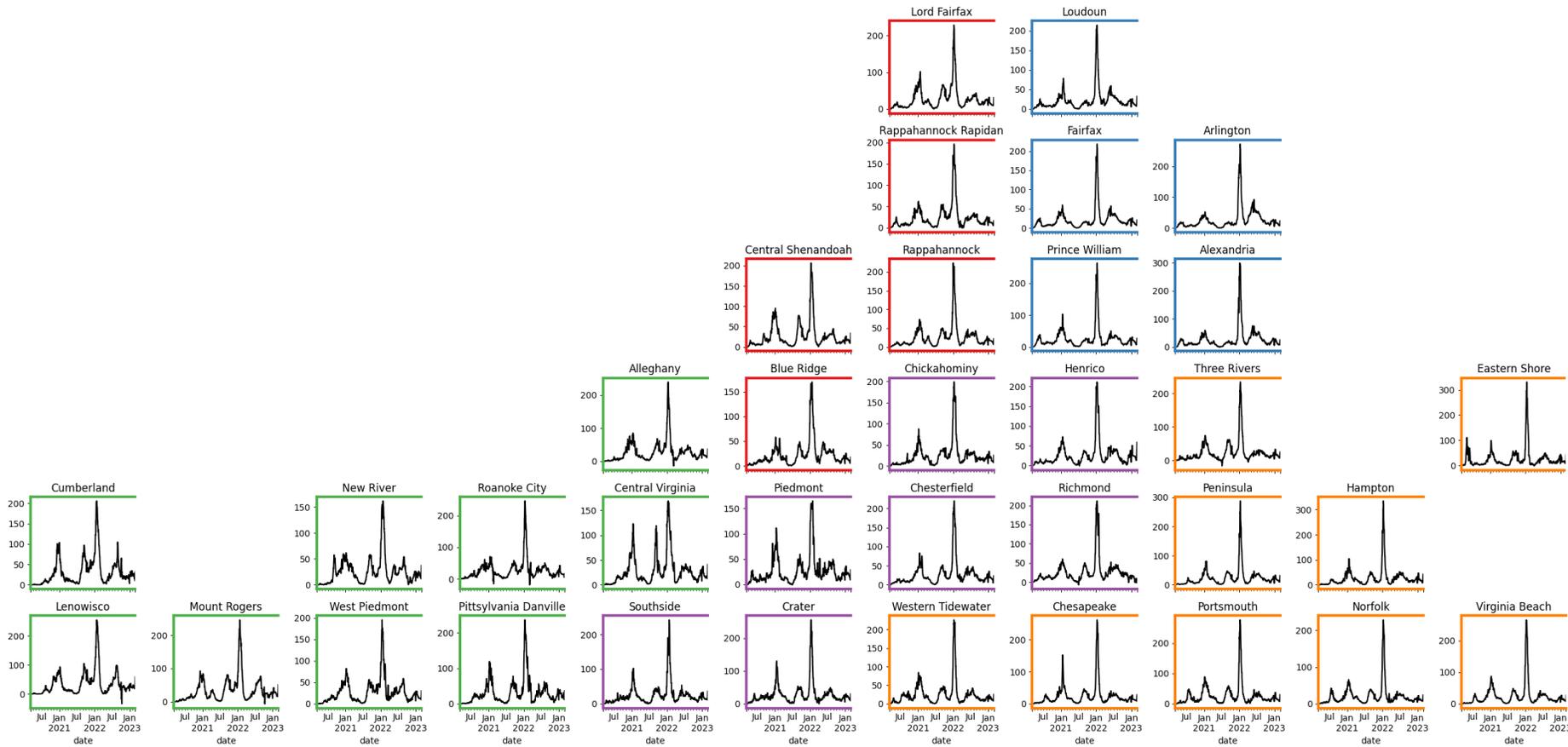
Even without perfect projections, we can confidently draw conclusions:

- Case rates and hospitalizations from COVID-19 continue to decline
- Case rates and hospitalizations from Influenza are very low
- **Model Updates**
  - Model updated to fit hospital admissions as opposed to cases
    - New technique used to determine hospitalization to infection ratio
  - Projection model updated this week, two non-specific scenarios related to increases in transmissibility.
  - Boosted transmissibility can generate new surge in activity which does not exceed levels from Summer of 2022

# COVID-19 Surveillance

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# Case Rates (per 100k) and Test Positivity

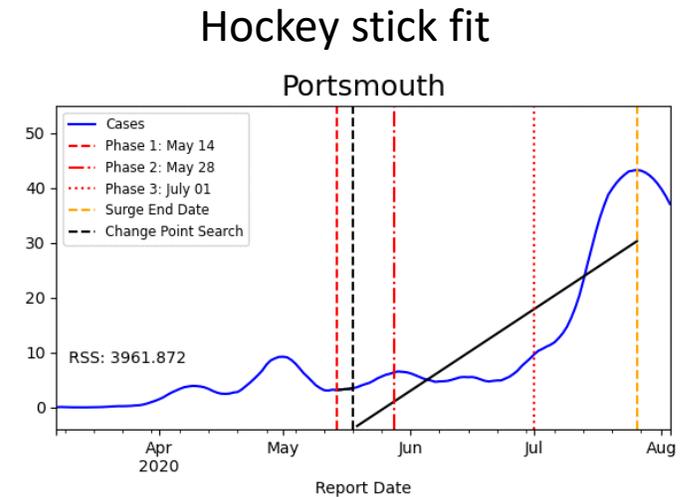


**County level RT-PCR test positivity**  
**Green:** <5.0% (or <20 tests in past 14 days)  
**Orange:** 5.0%-10.0% (or <500 tests and <2000 tests/100k and >10% positivity over 14 days)  
**Red:** >10.0% (and not "Green" or "Yellow")

# District Trajectories

**Goal:** Define epochs of a Health District's COVID-19 incidence to characterize the current trajectory

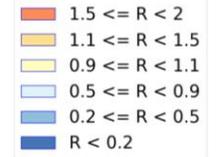
**Method:** Find recent peak and use hockey stick fit to find inflection point afterwards, then use this period's slope to define the trajectory



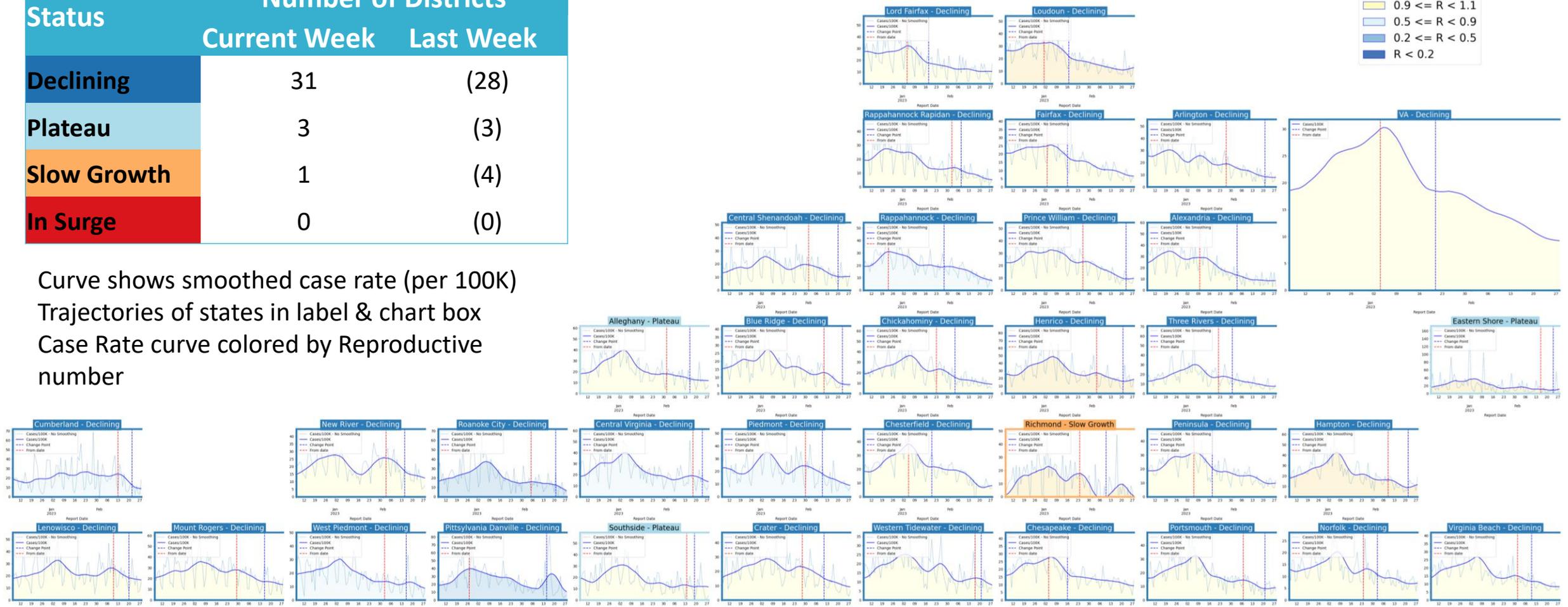
Trajectory	Description	Weekly Case Rate Slope (per 100k)	Weekly Hosp Rate Slope (per 100k)
<b>Declining</b>	Sustained decreases following a recent peak	slope < -0.88/day	slope < -0.07/day
<b>Plateau</b>	Steady level with minimal trend up or down	-0.88/day < slope < 0.42/day	-0.07/day < slope < 0.07/day
<b>Slow Growth</b>	Sustained growth not rapid enough to be considered a Surge	0.42/day < slope < 2.45/day	0.07/day < slope < 0.21/day
<b>In Surge</b>	Currently experiencing sustained rapid and significant growth	2.45/day < slope	0.21/day < slope

# District Case Trajectories – last 10 weeks

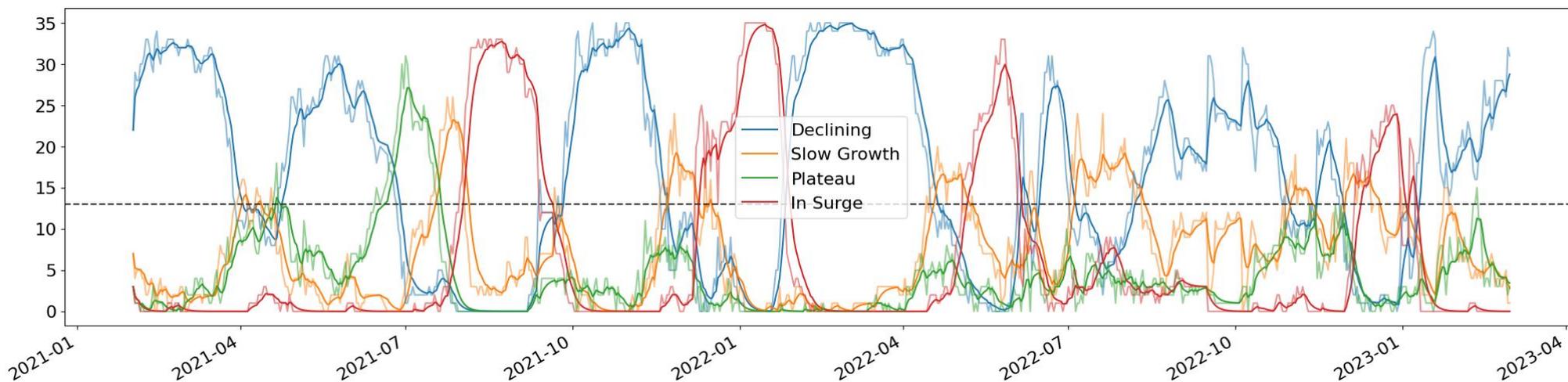
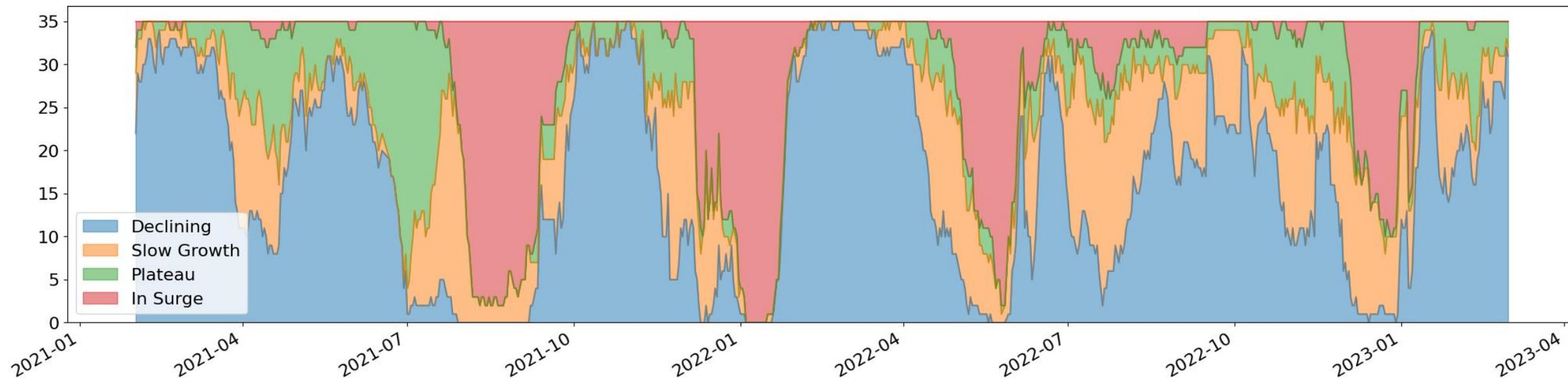
Status	Number of Districts	
	Current Week	Last Week
Declining	31	(28)
Plateau	3	(3)
Slow Growth	1	(4)
In Surge	0	(0)



Curve shows smoothed case rate (per 100K)  
 Trajectories of states in label & chart box  
 Case Rate curve colored by Reproductive number



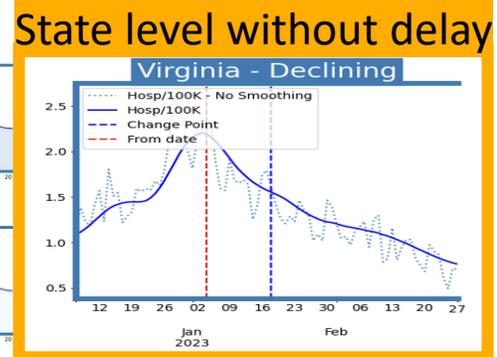
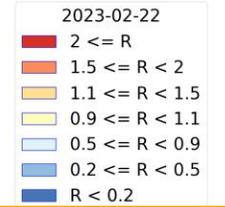
# District Case Trajectories – Full History



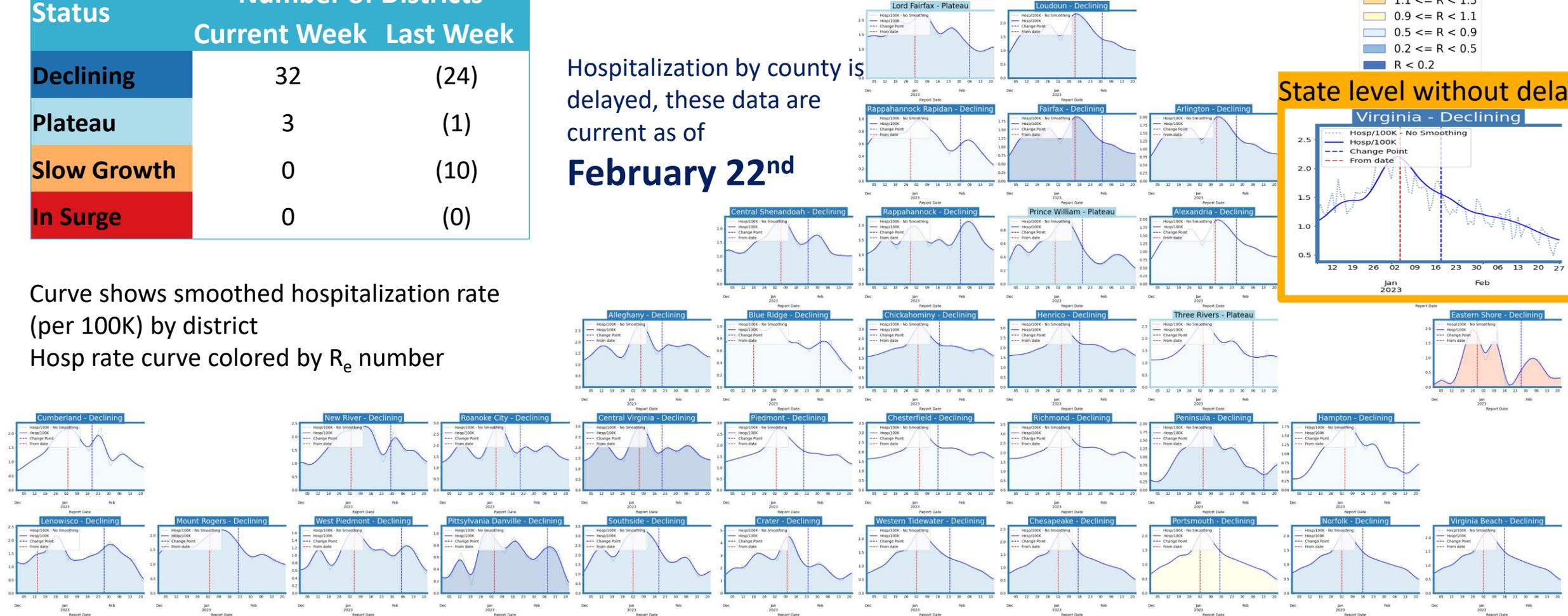
# District Hospital Trajectories – last 10 weeks

Status	Number of Districts	
	Current Week	Last Week
Declining	32	(24)
Plateau	3	(1)
Slow Growth	0	(10)
In Surge	0	(0)

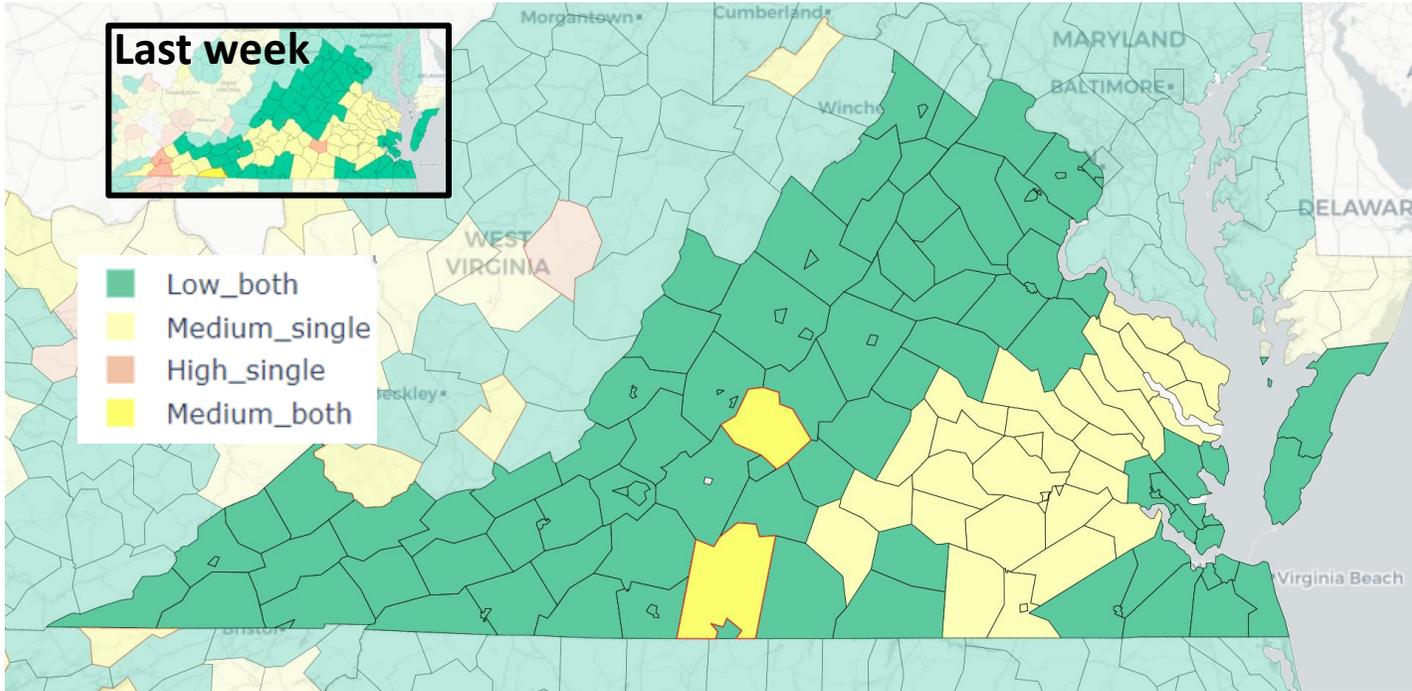
Hospitalization by county is delayed, these data are current as of **February 22<sup>nd</sup>**



Curve shows smoothed hospitalization rate (per 100K) by district  
Hosp rate curve colored by R<sub>e</sub> number



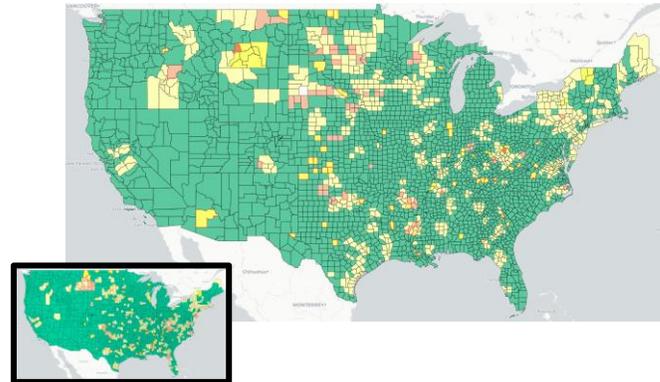
# CDC's COVID-19 Community Levels



**Red outline indicates county had 200 or more cases per 100k in last week**

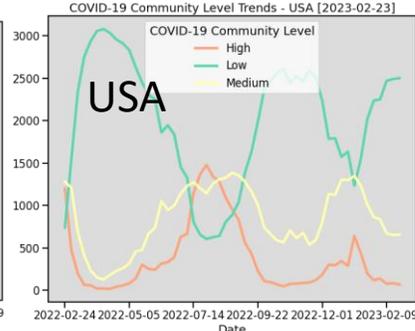
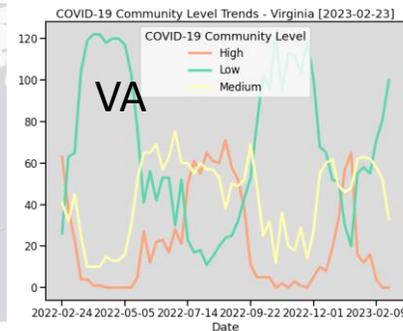
**Pale color indicates either beds or occupancy set the level for this county**

**Dark color indicates both beds and occupancy set the level for this county**



**Last week**

3-Mar-23

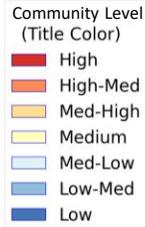


COVID-19 Community Levels - Use the Highest Level that Applies to Your Community				
New COVID-19 Cases Per 100,000 people in the past 7 days	Indicators	Low	Medium	High
Fewer than 200	New COVID-19 admissions per 100,000 population (7-day total)	<10.0	10.0-19.9	≥20.0
	Percent of staffed inpatient beds occupied by COVID-19 patients (7-day average)	<10.0%	10.0-14.9%	≥15.0%
200 or more	New COVID-19 admissions per 100,000 population (7-day total)	NA	<10.0	≥10.0
	Percent of staffed inpatient beds occupied by COVID-19 patients (7-day average)	NA	<10.0%	≥10.0%

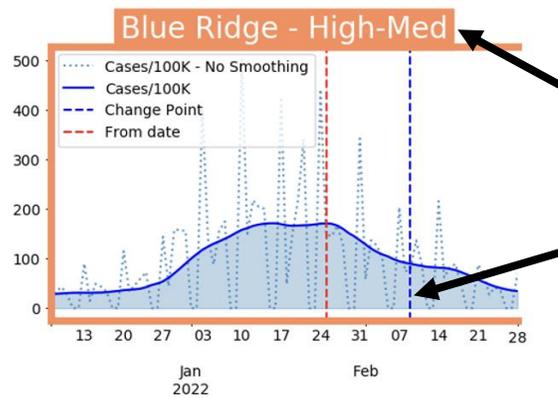
The COVID-19 community level is determined by the higher of the new admissions and inpatient beds metrics, based on the current level of new cases per 100,000 population in the past 7 days



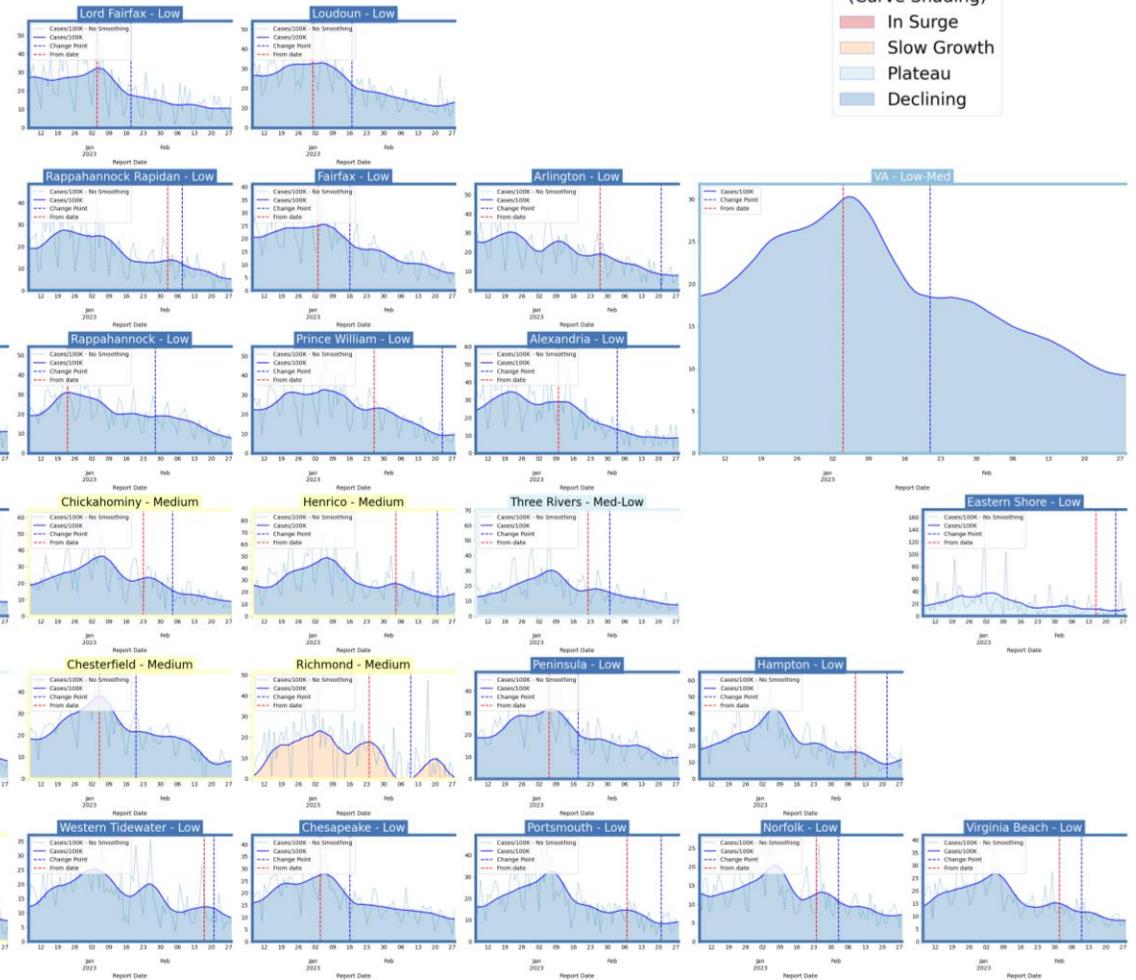
# District Trajectories with Community Levels



Curve shows smoothed case rate (per 100K)  
 CDC's new [Community Level](#) aggregated to district level in label & chart box color  
 Case Rate curve colored by Trajectory



District's Aggregate Community Level  
 Aggregate level a simple mean of all levels for counties in district  
 Case rate Trajectory



# COVID-19 Growth Metrics

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# Estimating Daily Reproductive Number – VDH report dates

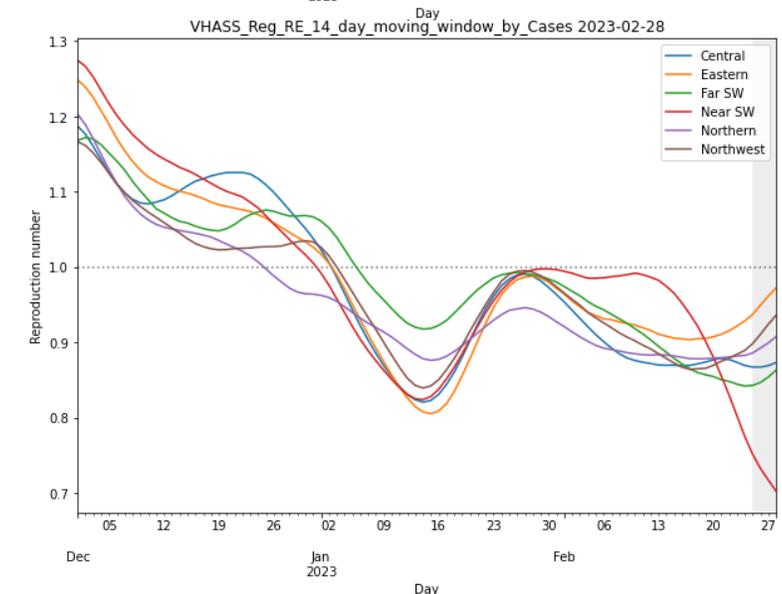
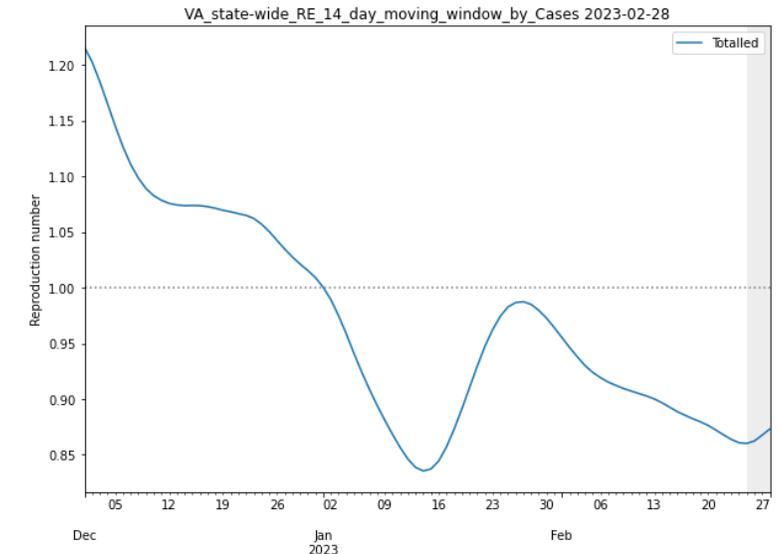
Feb 28<sup>th</sup> Estimates

Region	Date Confirmed $R_e$	Date Confirmed Diff Last Week
State-wide	0.873	0.069
Central	0.873	0.069
Eastern	0.972	0.173
Far SW	0.863	0.061
Near SW	0.702	-0.243
Northern	0.907	0.132
Northwest	0.936	0.160

## Methodology

- Wallinga-Teunis method (EpiEstim<sup>1</sup>) for cases by **confirmation date**
- Serial interval: updated to discrete distribution from observations (mean=4.3, Flaxman et al, Nature 2020)
- Using Confirmation date since due to increasingly unstable estimates from onset date due to backfill

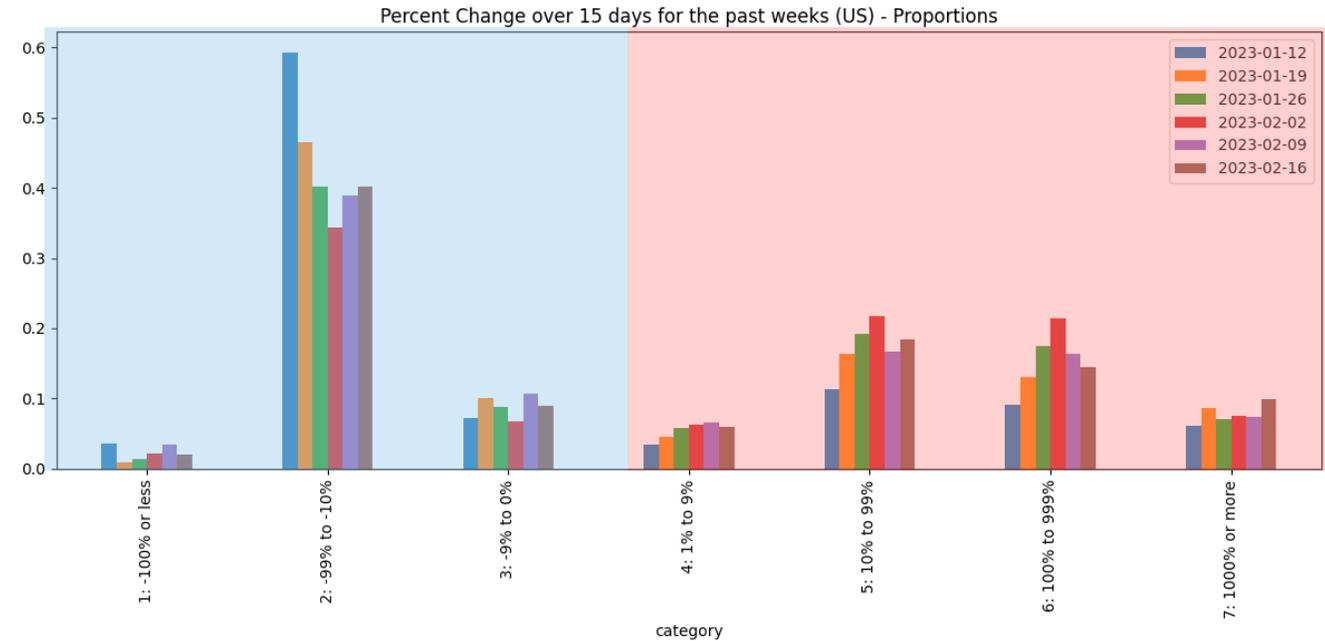
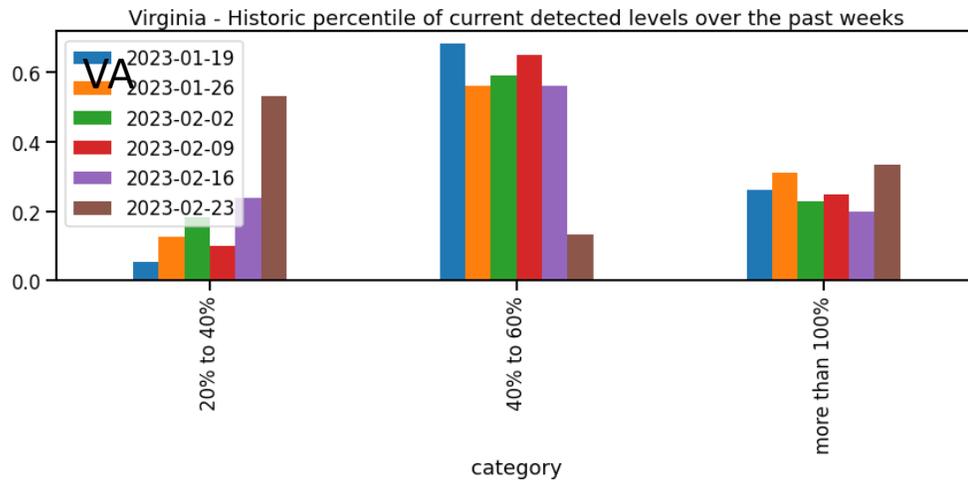
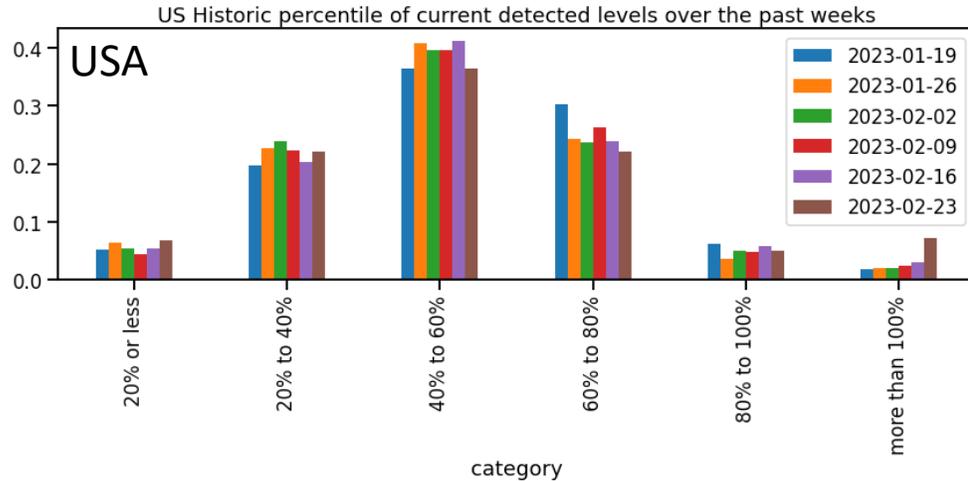
1. Anne Cori, Neil M. Ferguson, Christophe Fraser, Simon Cauchemez. A New Framework and Software to Estimate Time-Varying Reproduction Numbers During Epidemics. American Journal of Epidemiology, Volume 178, Issue 9, 1 November 2013, Pages 1505–1512, <https://doi.org/10.1093/aje/kwt133>



# Wastewater Monitoring

## Wastewater provides a coarse early warning of COVID-19 levels in communities

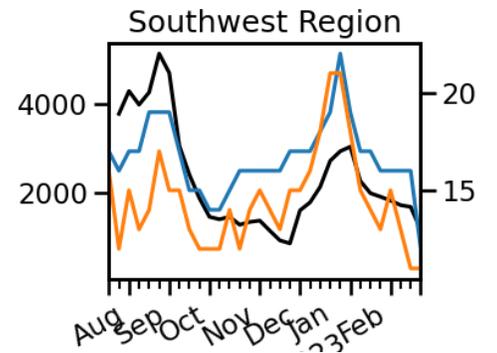
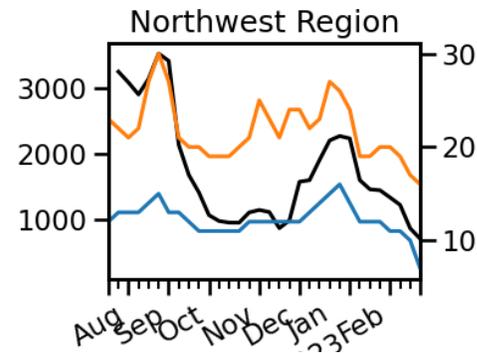
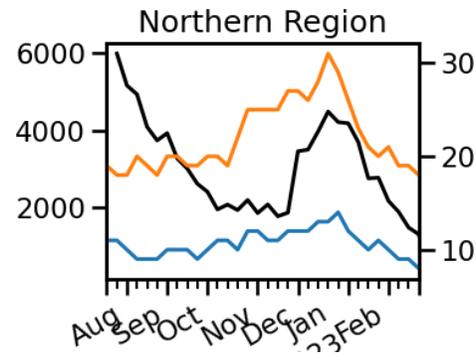
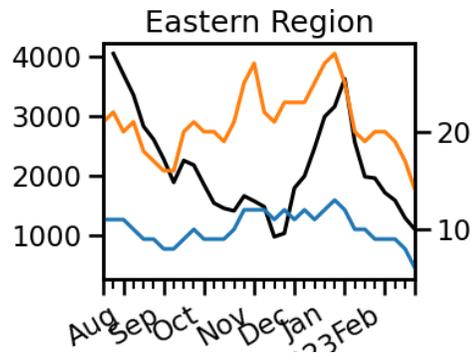
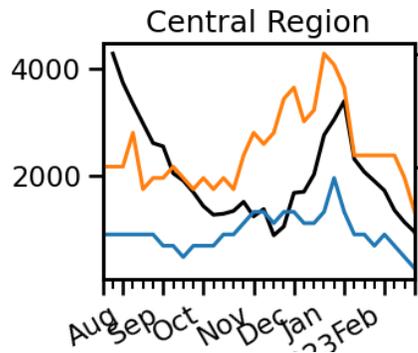
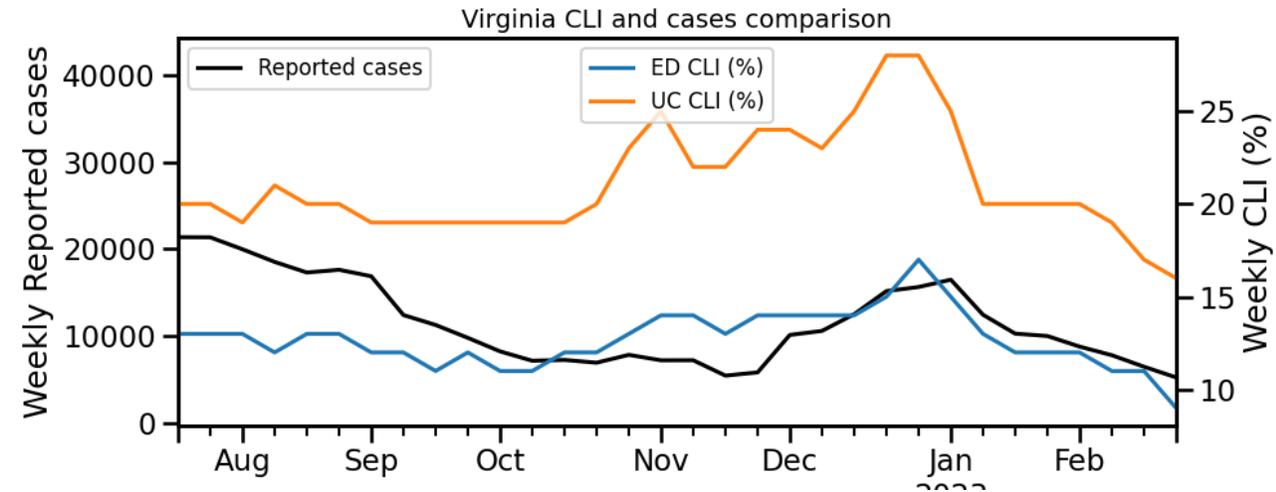
- Overall in the US, there is an increase in sites with increased levels of virus compared to 15 days ago
- Growth seen in the category where current virus levels are at or exceeding max of previous historical levels



# COVID-like Illness Activity

## COVID-like Illness (CLI) gives a measure of COVID transmission in the community

- Emergency Dept (ED) based CLI is more correlated with case reporting
- Urgent Care (UC) is a leading indicator but may be influenced by testing for other URIs
- **After recent surges, levels are now at lowest levels in past 7 months**



# COVID-19 Severity Metrics

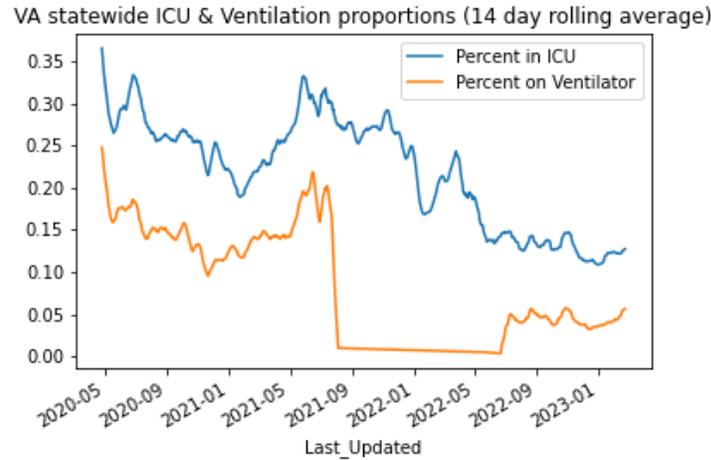
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# Hospitalizations and Severe Outcomes

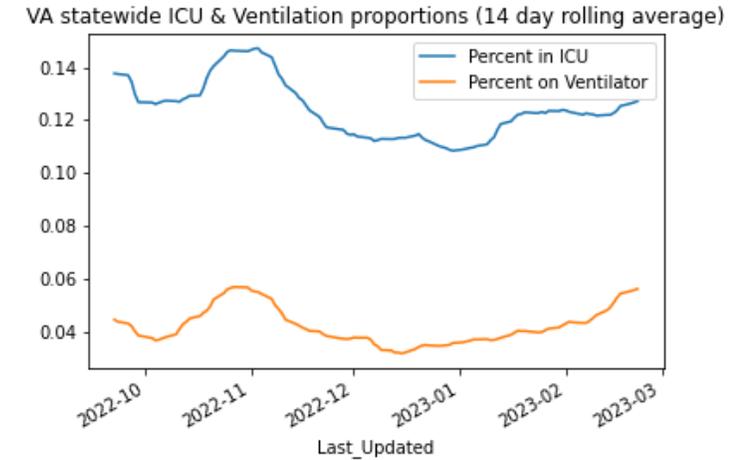
## Proportion of most severe outcomes decreasing among those who are hospitalized

- ICU has declined from ~20% of hospitalized to 10-15% since initial Omicron wave
- Recent trend tipping up, though current levels near historic lows
- Regional variation tracks state-level

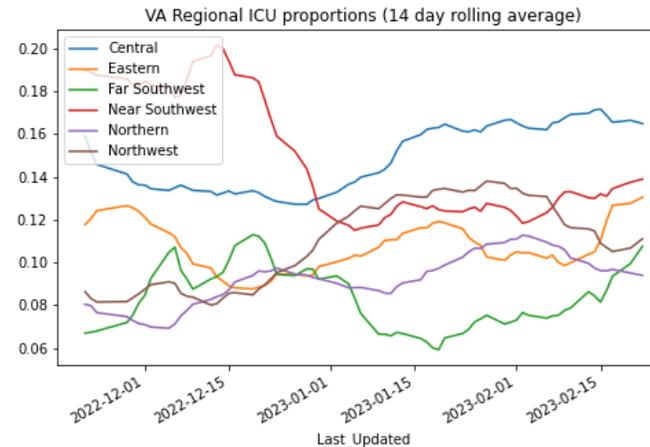
Virginia-wide – full pandemic



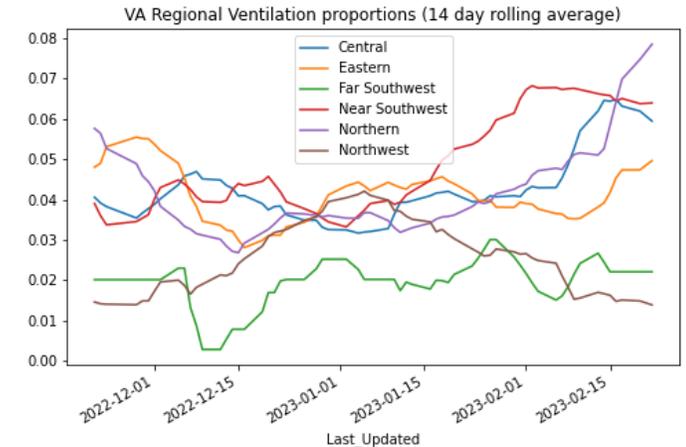
Virginia-wide – recent



Virginia Regional ICU percent



Virginia Regional Ventilation %



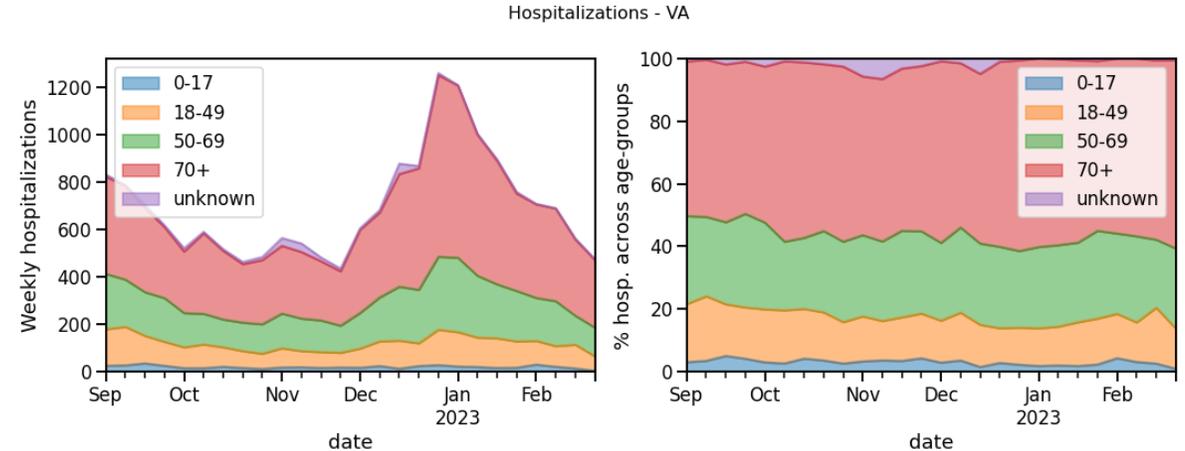
# Hospitalizations in VA by Age

## Age distribution in hospitals relatively stable

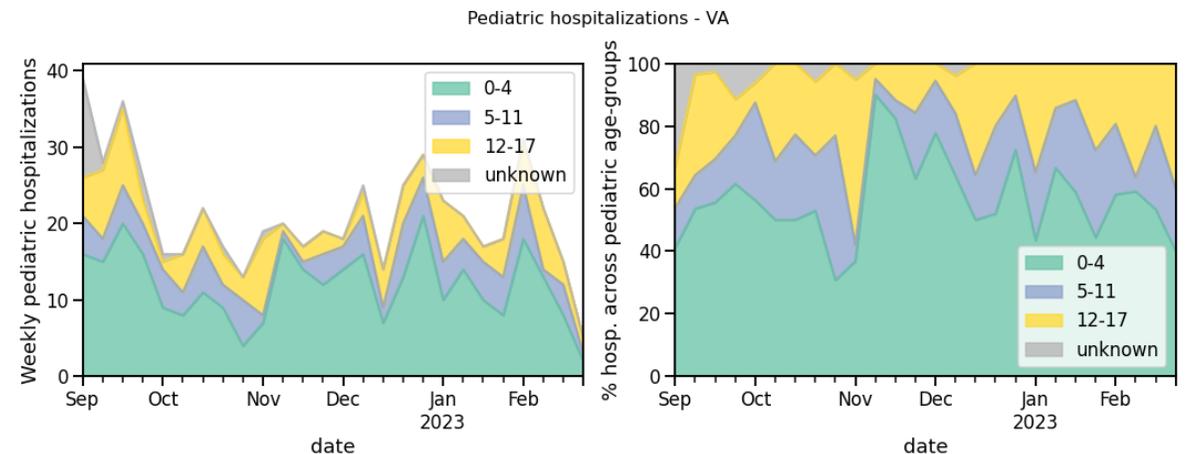
- Uptick in hospitalizations mostly fueled by 70+ age group
- Pediatric hospitalizations declining

Note: These data are lagged and based on HHS hospital reporting

## Virginia Hospitalizations by Age (all ages)



## Pediatric Hospitalizations by Age (0-17yo)



# COVID-19 Spatial Epidemiology

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# Zip code level weekly Case Rate (per 100K)

## Case Rates in the last week by zip code

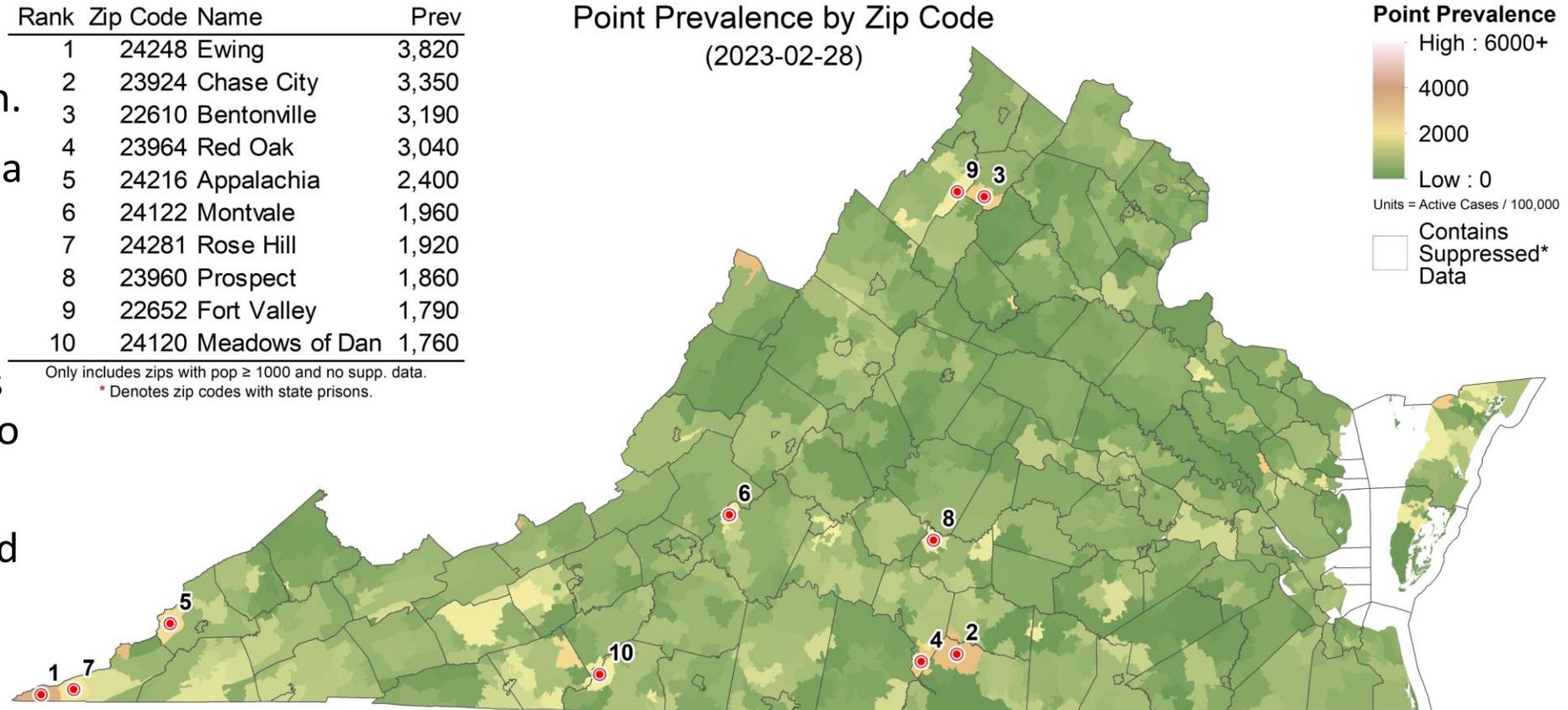
- Statewide prevalence peaked in early January. Overall case rates have declined 48% in a month.
- No zip codes containing a prison appear in this week's top 10.
- Though there are a handful of small clusters in rural regions, all metro areas are in the "green".
- Some counts are low and suppressed to protect anonymity. They are shown with a red outline.

Rank	Zip Code	Name	Prev
1	24248	Ewing	3,820
2	23924	Chase City	3,350
3	22610	Bentonville	3,190
4	23964	Red Oak	3,040
5	24216	Appalachia	2,400
6	24122	Montvale	1,960
7	24281	Rose Hill	1,920
8	23960	Prospect	1,860
9	22652	Fort Valley	1,790
10	24120	Meadows of Dan	1,760

Only includes zips with pop ≥ 1000 and no supp. data.

\* Denotes zip codes with state prisons.

Point Prevalence by Zip Code  
(2023-02-28)



Based on Spatial Empirical Bayes smoothed point prevalence, with an 8:1 ascertainment ratio, for week ending 2023-02-28.

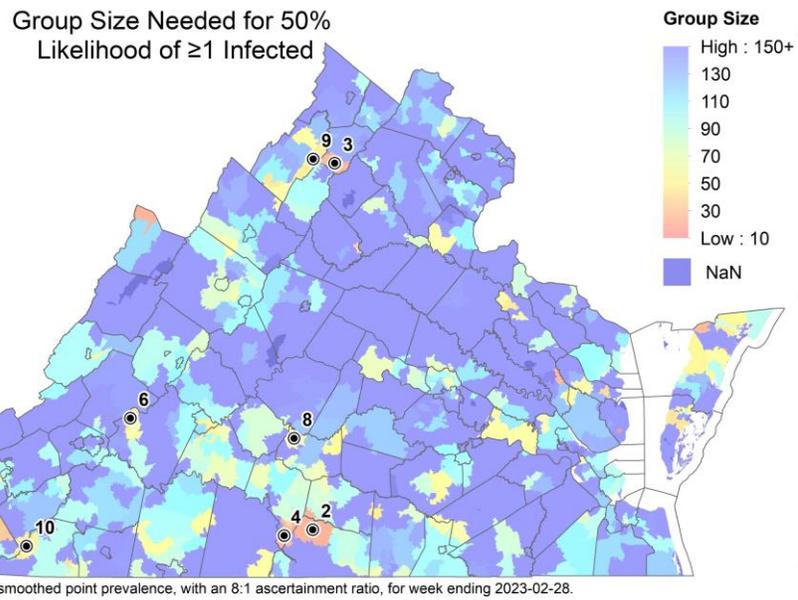
# Risk of Exposure by Group Size and HCW prevalence

## Case Prevalence in the last week by zip code used to calculate risk of encountering someone infected in a gathering of randomly selected people

- **Group Size:** Assumes **8 undetected infections** per confirmed case (ascertainment rate from recent seroprevalence survey) and shows minimum size of a group with a 50% chance an individual is infected by zip code (e.g., in a group of 18 in Ewing, there is a 50% chance someone will be infected).
- **HCW ratio:** Case rate among health care workers (HCW) in the last week using patient facing health care workers as the numerator / population's case prevalence. Most highlighted counties have < 3 HCW cases.

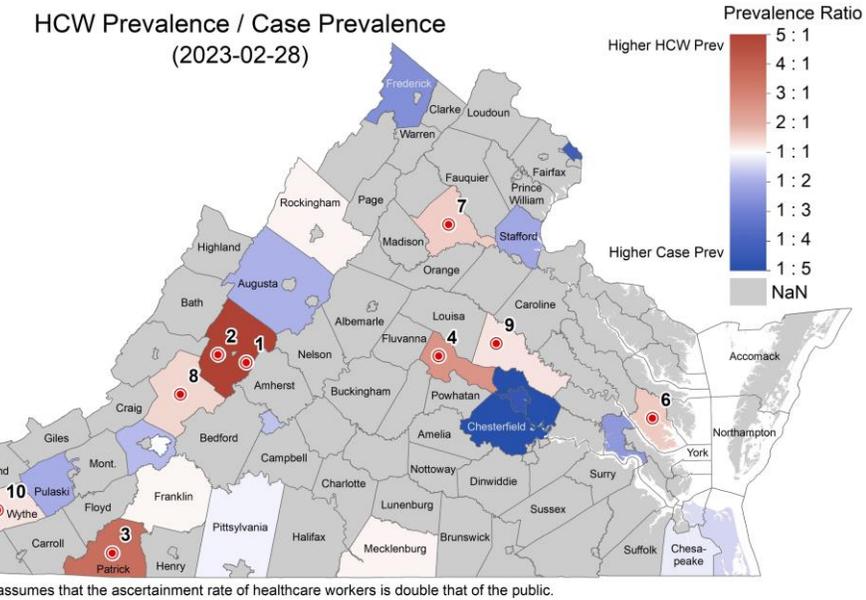
Rank	Zip Code	Name	Size
1	24248	Ewing	18
2	23924	Chase City	20
3	22610	Bentonville	21
4	23964	Red Oak	22
5	24216	Appalachia	29
6	24122	Montvale	35
7	24281	Rose Hill	36
8	23960	Prospect	37
9	22652	Fort Valley	38
10	24120	Meadows of Dan	39

Only includes zip codes with pop ≥ 1000 and no supp. data.  
 \* Denotes zip codes with state prisons.



Rank	Name	Ratio
1	Buena Vista City	5.1
2	Rockbridge County	5.0
3	Patrick County	3.6
4	Goochland County	2.6
5	Smyth County	2.1
6	Gloucester County	1.6
7	Culpeper County	1.6
8	Botetourt County	1.5
9	Hanover County	1.3
10	Wythe County	1.3

Beyond Scale



Based on Spatial Empirical Bayes smoothed point prevalence, with an 8:1 ascertainment ratio, for week ending 2023-02-28.

# Current Hot-Spots

## Case rates that are significantly different from neighboring areas or model projections

- **Spatial:** Getis-Ord  $G_i^*$  based hot spots compare clusters of zip codes with weekly case prevalence higher than nearby zip codes to identify larger areas with statistically significant deviations
- **Temporal:** The weekly case rate (per 100K) projected last month compared to those observed by county, which highlights temporal fluctuations that differ from the model's projections.
- Hot spots are sporadic around the state. Models overpredicted New River and Lenowisco. Piedmont also showed slightly fewer cases than expected. The rest of the Commonwealth tracked the model forecasts.

### Spatial Hotspots

Spot	Zip Code	Name	Conf.
1	24248	Ewing	99%
2	23924	Chase City	99%
3	22610	Bentonville	99%
4	23964	Red Oak	99%
5	24216	Appalachia	99%

Only zips with pop  $\geq 1000$  and no supp. data.  
 \* Denotes zip codes with state prisons.

Point Prevalence Hot Spots by Zip Code  
(2023-02-28)

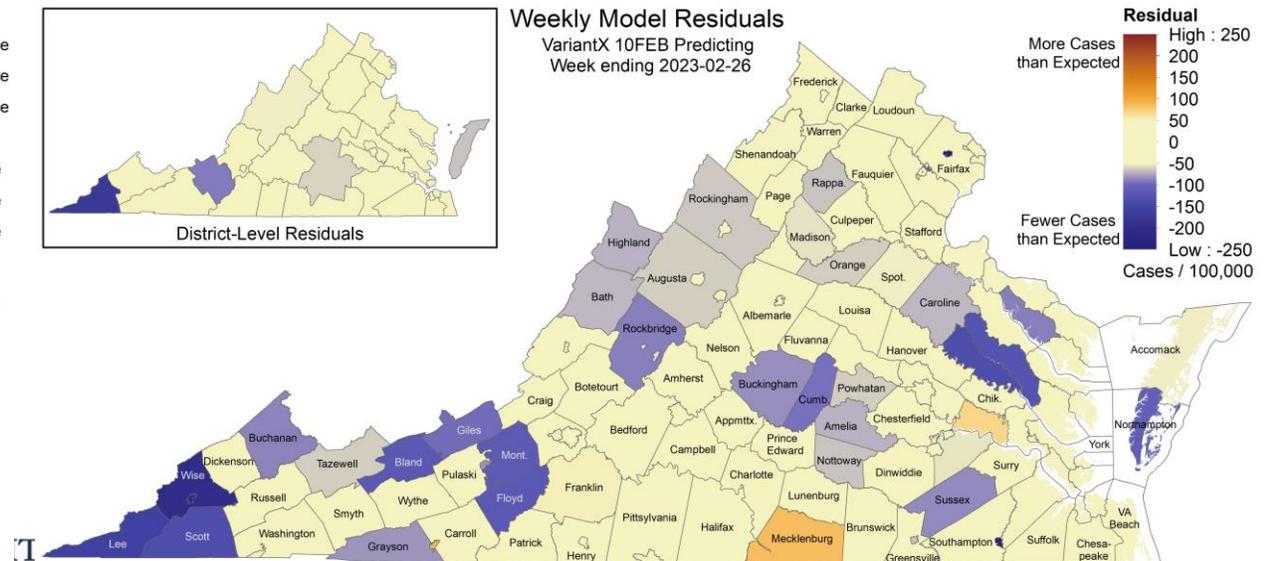


**Getis-Ord  $G_i^*$  HotSpots**

- Cold Spot - 99% Confidence
- Cold Spot - 95% Confidence
- Cold Spot - 90% Confidence
- Not Significant
- Hot Spot - 90% Confidence
- Hot Spot - 95% Confidence
- Hot Spot - 99% Confidence

### Clustered Temporal Hotspots

Weekly Model Residuals  
 VariantX 10FEB Predicting  
 Week ending 2023-02-26

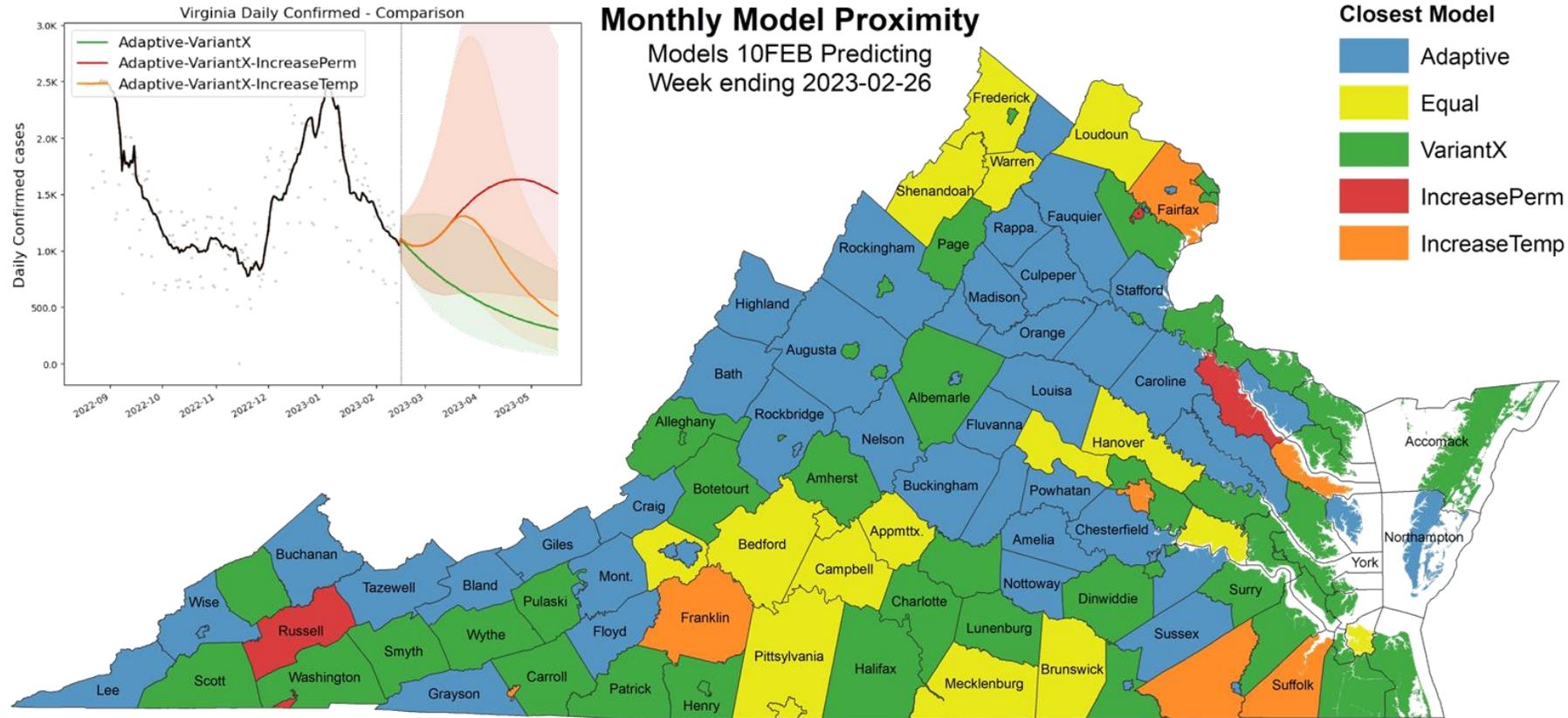


Based on Global Empirical Bayes smoothed point prevalence for week ending 2023-02-28.

3-Mar-23

# Scenario Trajectory Tracking

Which scenario best tracked the ground truth for each county?

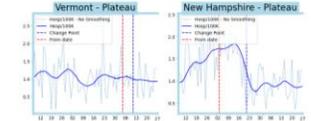
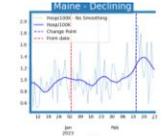
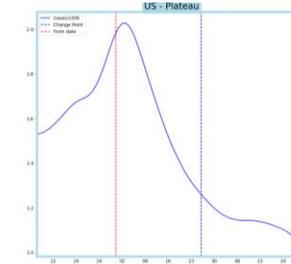
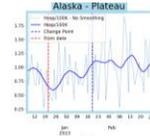


- Statewide cases fell between Adaptive and VariantX. The former underpredicted cases, and the latter overpredicted. Counties were evenly split between the two. Only a few counties followed the projections of the IncreasePerm and IncreasedTemp scenarios.

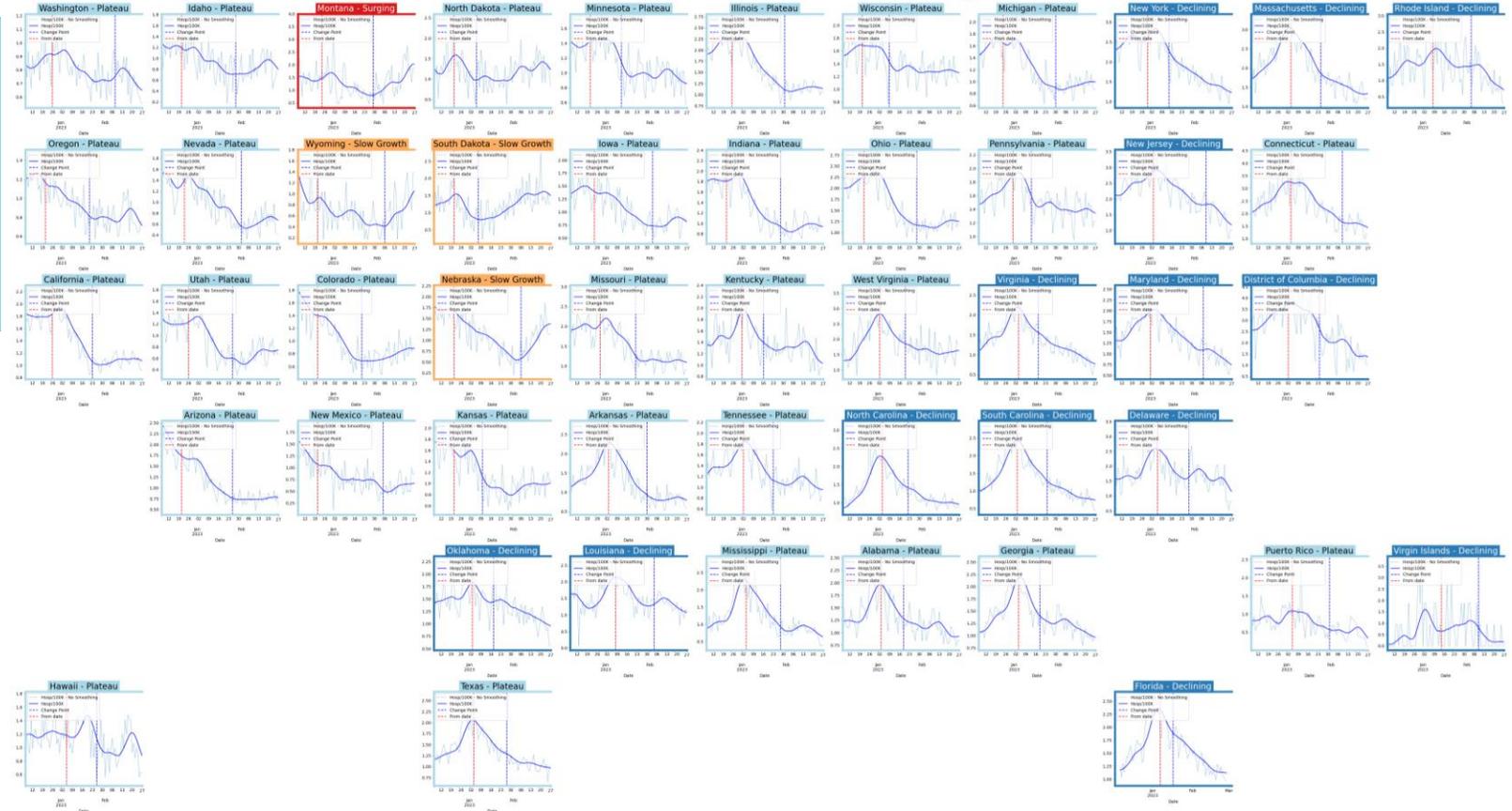
# COVID-19 Broader Context

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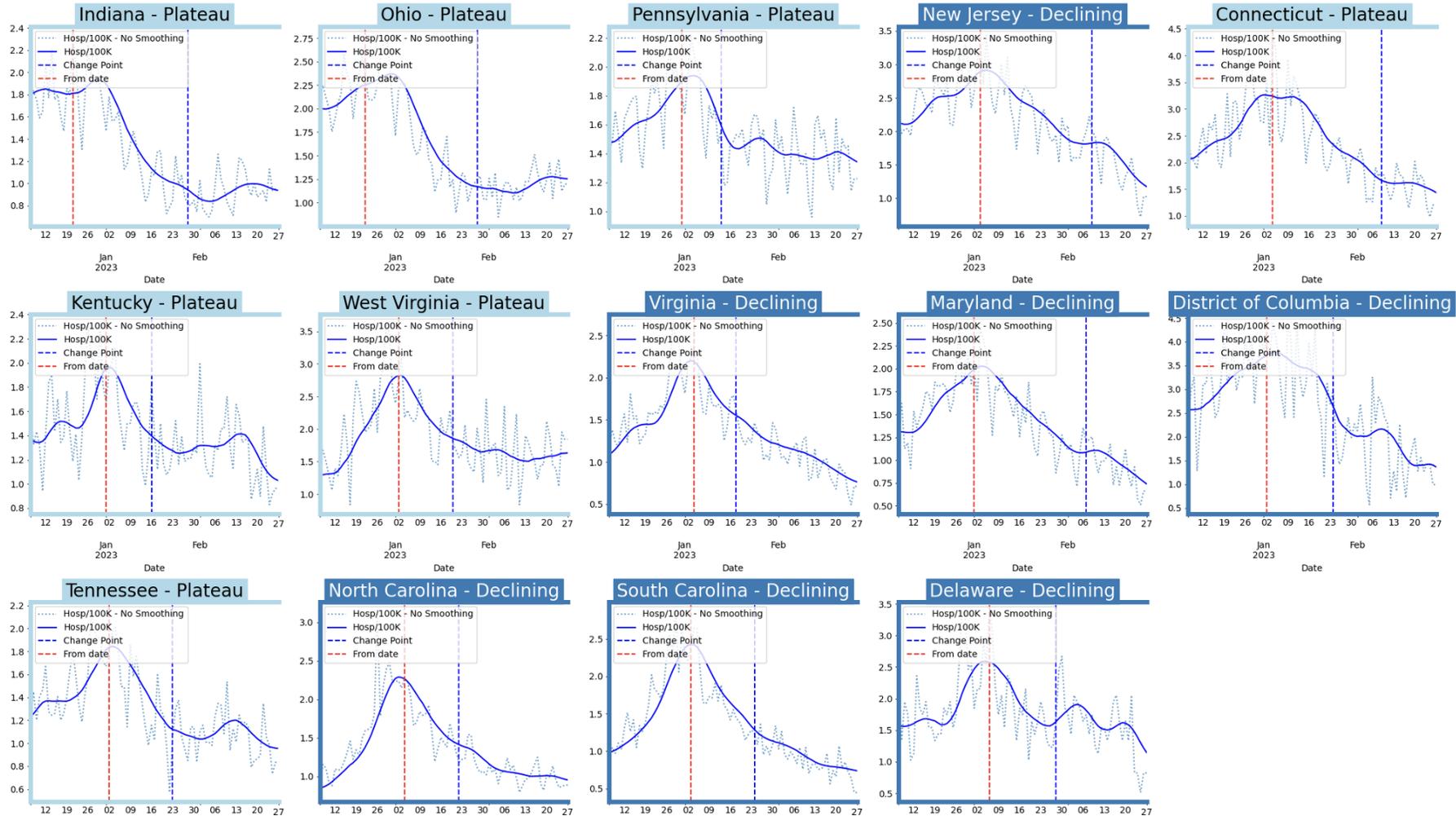
# United States Hospitalizations



Status	Number of States Current Week	Last Week
Declining	15	(14)
Plateau	34	(28)
Slow Growth	3	(11)
In Surge	1	(0)



# Virginia and Her Neighbors

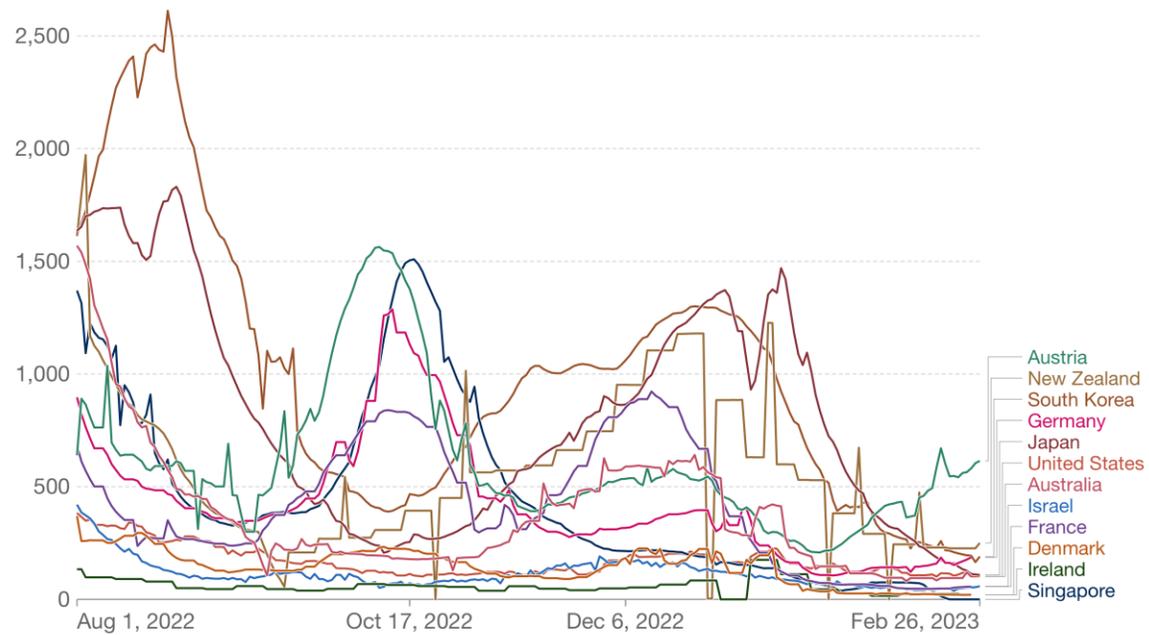


# Around the World – Various trajectories

## Confirmed cases

### Daily new confirmed COVID-19 cases per million people

7-day rolling average. Due to limited testing, the number of confirmed cases is lower than the true number of infections.



Source: Johns Hopkins University CSSE COVID-19 Data

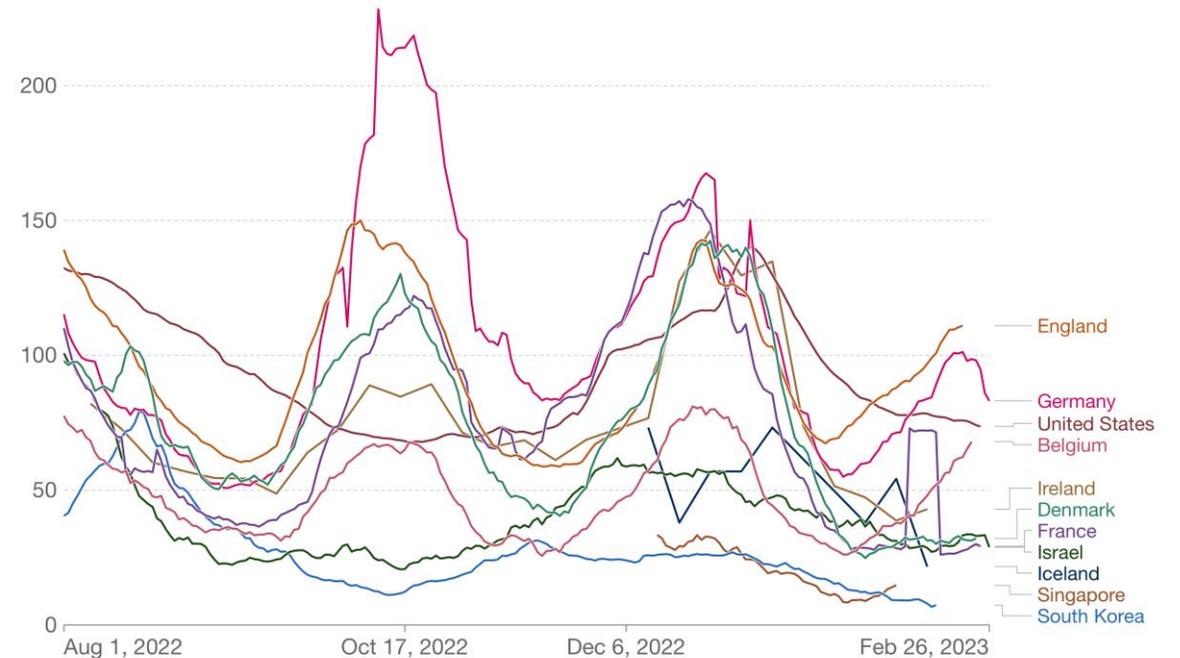
Our World in Data

CC BY

## Hospitalizations

### Weekly new hospital admissions for COVID-19 per million people

Weekly admissions refer to the cumulative number of new admissions over the previous week.



Source: Official data collated by Our World in Data

Our World in Data

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Our World in Data

[Our World in Data](https://ourworldindata.org)

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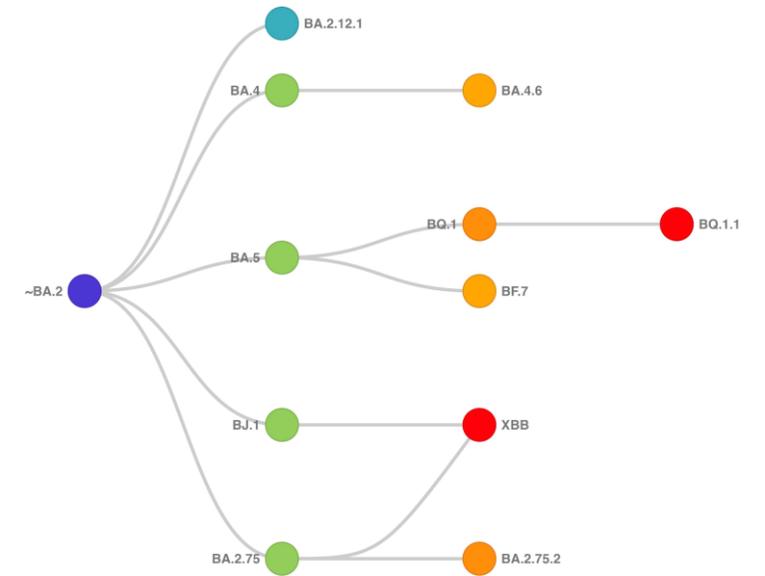
# COVID-19 Genomic Update

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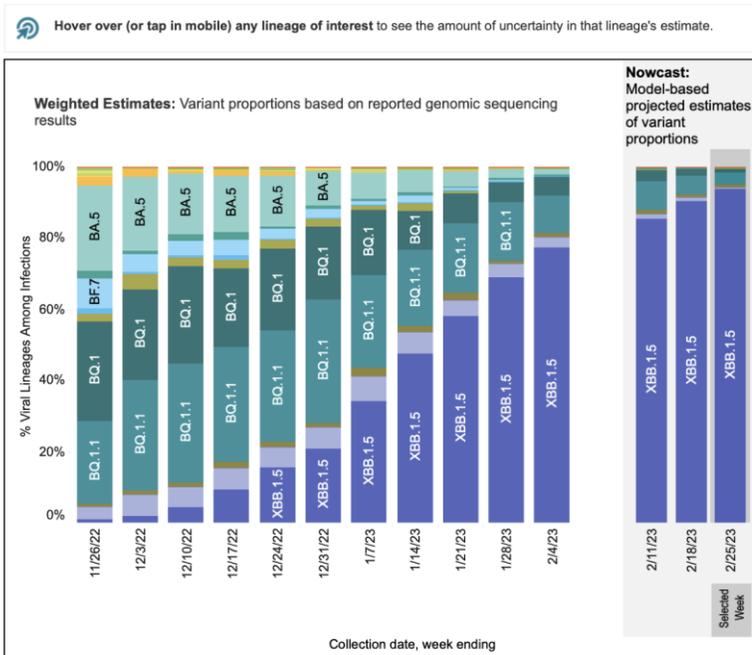
# SARS-CoV2 Variants of Concern

Emerging variants have potential to continue to alter the future trajectories of pandemic and have implications for future control

- Variants have been observed to: increase transmissibility, increase severity (more hospitalizations and/or deaths), and limit immunity provided by prior infection and vaccinations



Weighted and Nowcast Estimates in HHS Region 3 for Weeks of 11/20/2022 – 2/25/2023



Nowcast Estimates in HHS Region 3 for 2/19/2023 – 2/25/2023

Region 3 - Delaware, District of Columbia, Maryland, Pennsylvania, Virginia, and West Virginia

WHO label	Lineage #	US Class	%Total	95%PI
Omicron	XBB.1.5	VOC	93.7%	92.3-94.9%
	BQ.1.1	VOC	3.6%	3.0-4.4%
	BQ.1	VOC	1.1%	0.9-1.5%
	XBB	VOC	0.7%	0.6-0.9%
	CH.1.1	VOC	0.5%	0.3-0.8%
	BN.1	VOC	0.1%	0.1-0.2%
	BA.5	VOC	0.0%	0.0-0.1%
	BF.7	VOC	0.0%	0.0-0.1%
	BA.5.2.6	VOC	0.0%	0.0-0.0%
	BA.2	VOC	0.0%	0.0-0.0%
	BF.11	VOC	0.0%	0.0-0.0%
	BA.2.75	VOC	0.0%	0.0-0.0%
	BA.2.75.2	VOC	0.0%	0.0-0.0%
	BA.4.6	VOC	0.0%	0.0-0.0%
	B.1.1.529	VOC	0.0%	0.0-0.0%
	BA.4	VOC	0.0%	0.0-0.0%
	BA.2.12.1	VOC	0.0%	0.0-0.0%
	BA.1.1	VOC	0.0%	0.0-0.0%
Delta	B.1.617.2	VBM	0.0%	0.0-0.0%
Other	Other*		0.0%	0.0-0.1%

<https://clades.nextstrain.org>

## Omicron Updates\*

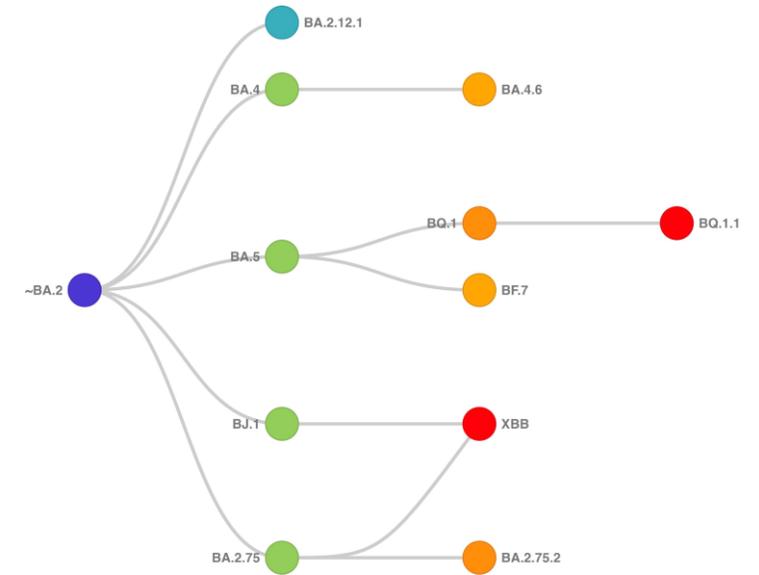
- XBB.1.5 has grown rapidly now accounting for 91%
- BQ.1 and BQ.1.1 continue to lose ground at 2% and 5% respectively
- XBB not in XBB.1.5 has fallen 1%
- BN.1 separated from BA.2.75 has fallen to 0.2%
- CH.1.1 has fallen from 2% during the holidays to just 0.7% now

\*percentages are CDC NowCast Estimates

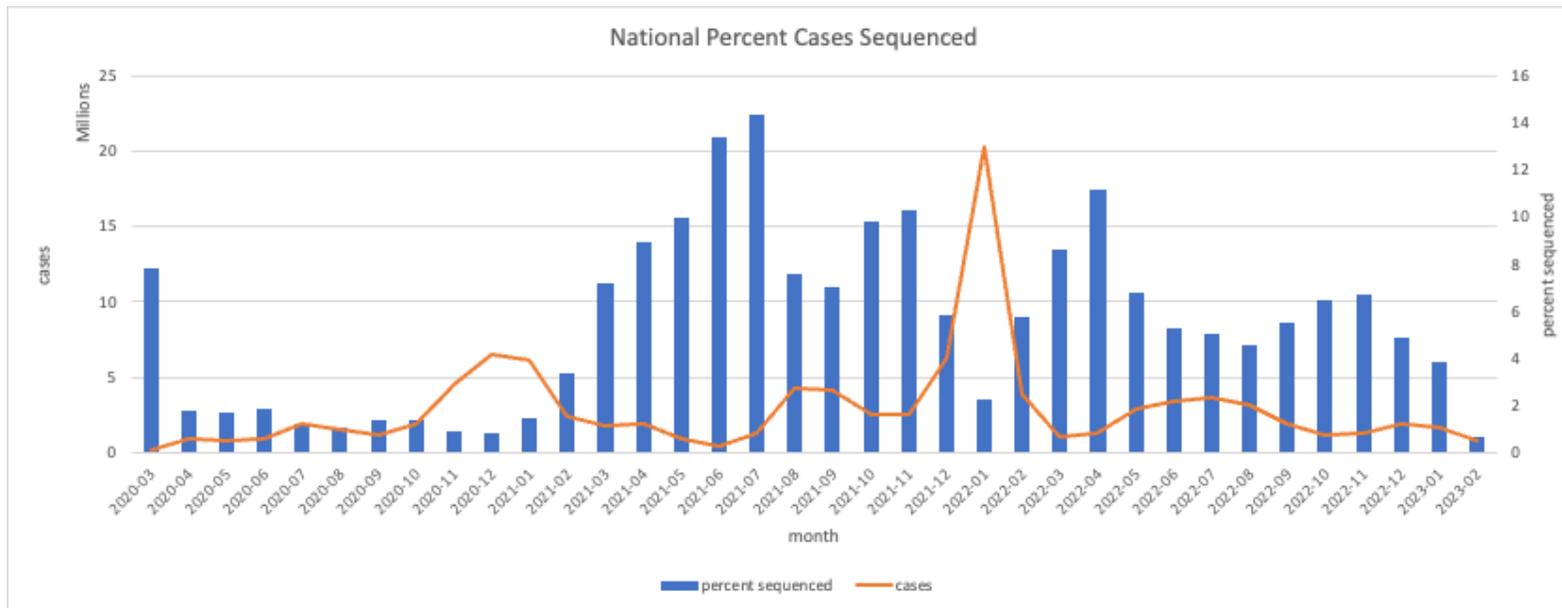
# SARS-CoV2 Sequencing

**Emerging variants have potential to continue to alter the future trajectories of pandemic and have implications for future control**

- Current proportion of cases being sequenced is on a downward trend nationally.
- Leveraging additional resources such as wastewater sequencing and adopting into existing infrastructure will be an important supplement



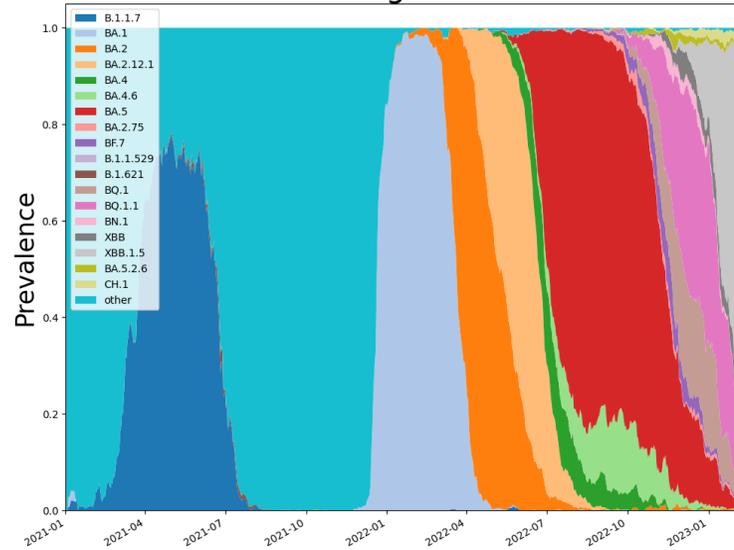
<https://clades.nextstrain.org>



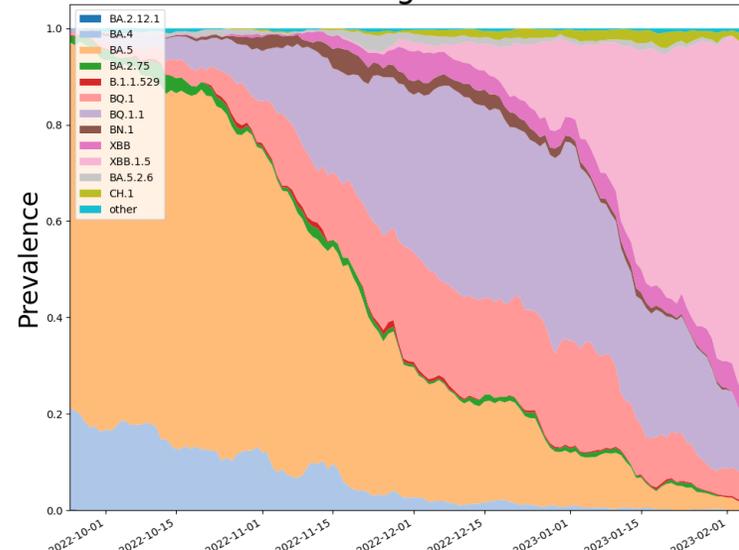
# SARS-CoV2 Omicron Sub-Variants

As detected in whole Genomes in public repositories

Virginia

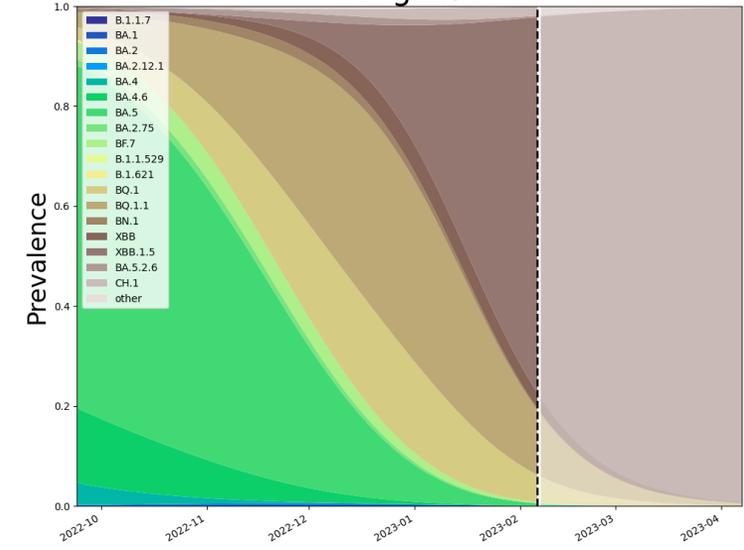


Virginia

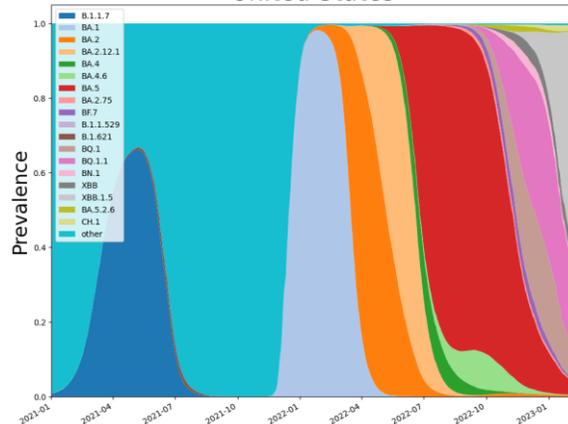


VoC Polynomial Fit Projections

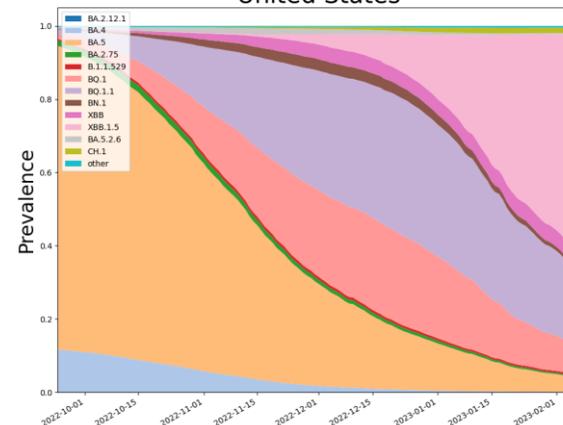
Virginia



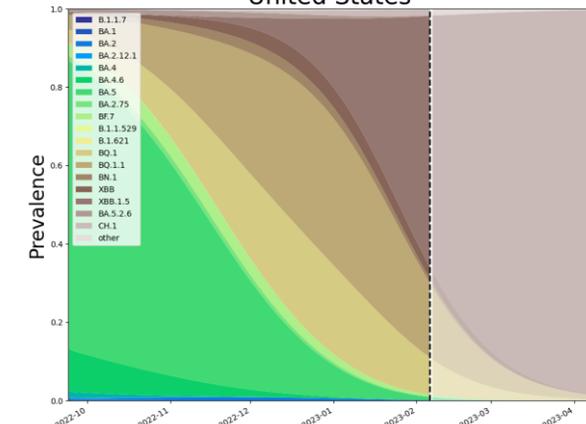
United States



United States



United States



Note:  
Everything from dotted line forward is a projection.

# SARS-CoV2 Omicron Sub-Variants

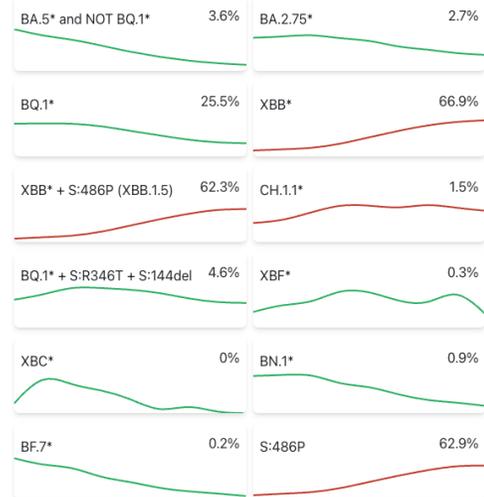
## COV-spectrum

“Editor’s choice”  
Variants to watch

### National

Which variant would you like to explore?

Editor's choice ▾



covSPECTRUM

Enabled by data from 

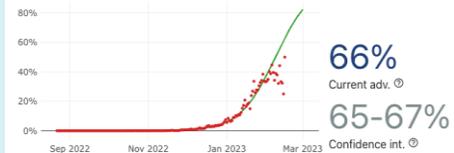
3-Mar-23

## XBB.1.5

### Relative growth advantage

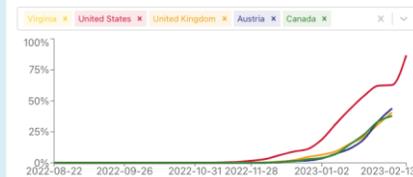
If variants spread pre-dominantly by local transmission across demographic group... (show more)

Estimated proportion through time

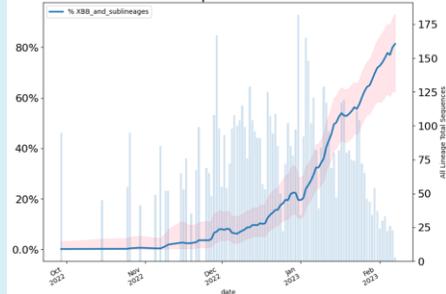


(\*) Assumes that the current advantage is due to an intrinsic viral advantage (a combination of increased transmission, immune escape, and prolonged infectious period).

### International comparison



### Virginia - 81.4% (XBB and sublineages) Last Sample: 2023-02-07



## BA.2.75.\*

### Relative growth advantage

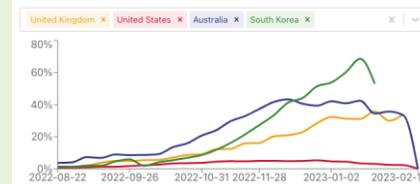
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Estimated proportion through time

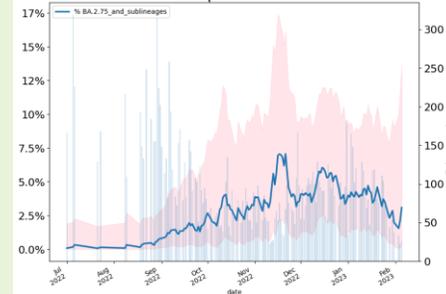


(\*) Assumes that the current advantage is due to an intrinsic viral advantage (a combination of increased transmission, immune escape, and prolonged infectious period).

### International comparison



### Virginia - 3.1% (BA.2.75 and sublineages) Last Sample: 2023-02-05

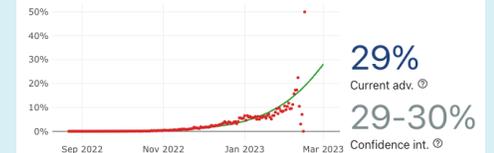


## CH.1.1\*

### Relative growth advantage

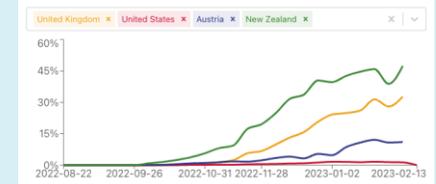
If variants spread pre-dominantly by local transmission across demographic group... (show more)

Estimated proportion through time

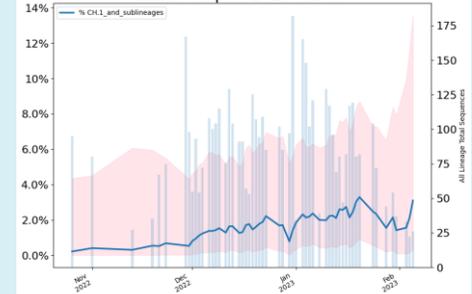


(\*) Assumes that the current advantage is due to an intrinsic viral advantage (a combination of increased transmission, immune escape, and prolonged infectious period).

### International comparison



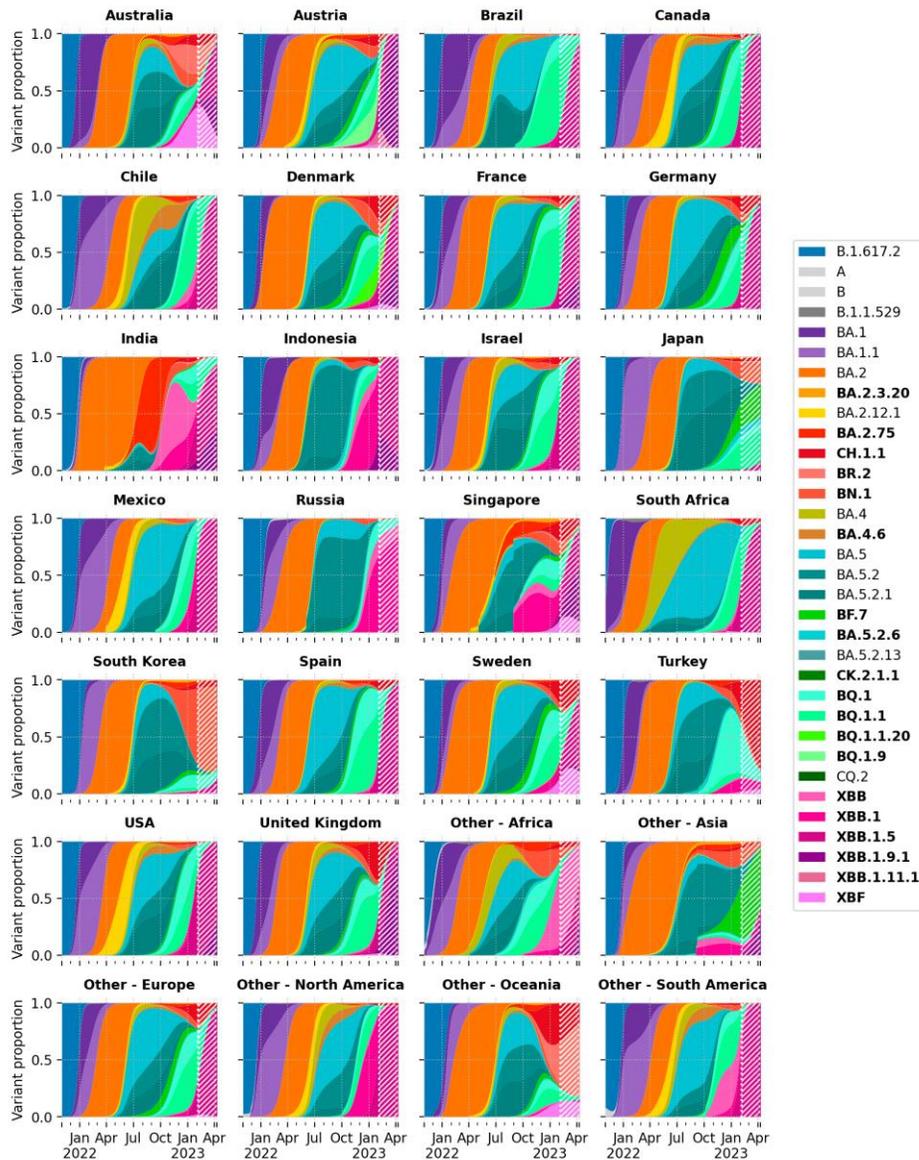
### Virginia - 3.1% (CH.1 and sublineages) Last Sample: 2023-02-05



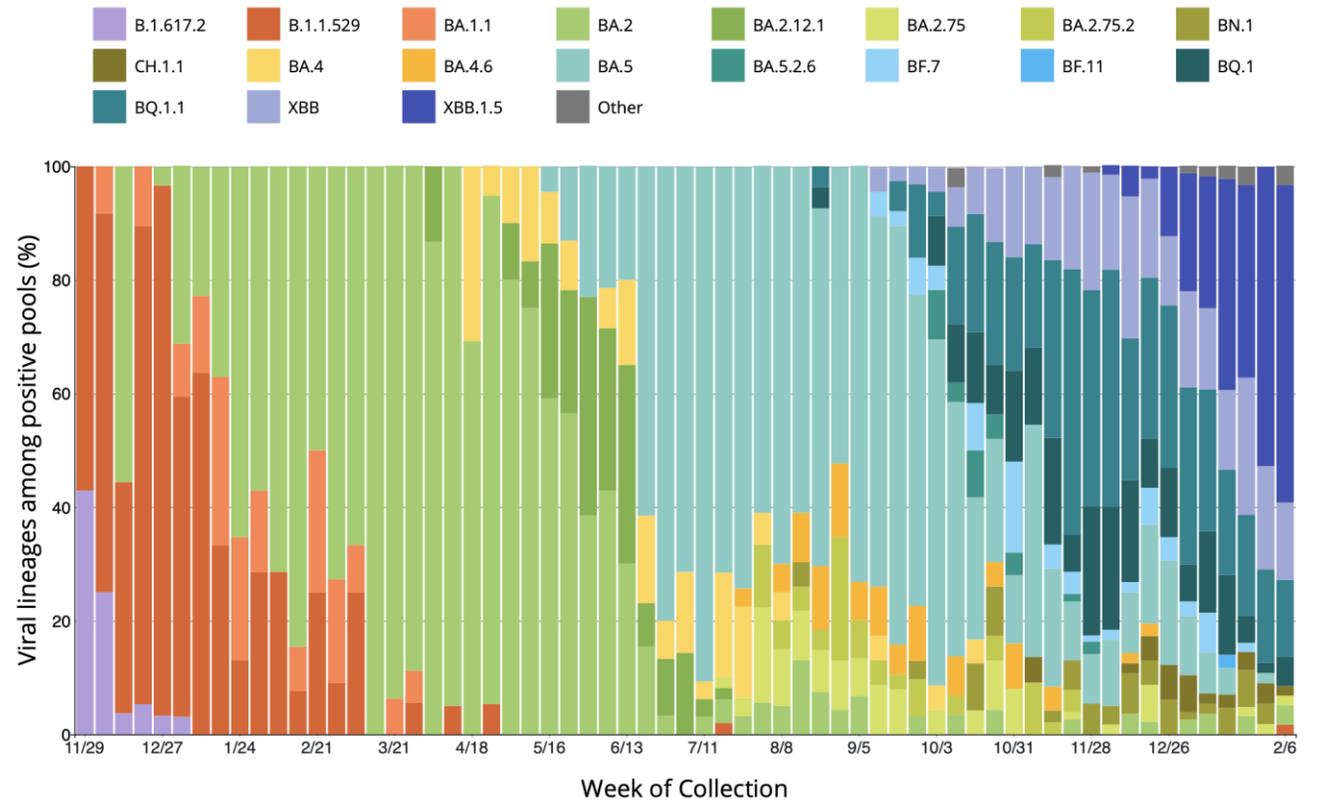
 UNIVERSITY of VIRGINIA

BIOCOMPLEXITY INSTITUTE

# Global SARS-CoV-2 Variant Status



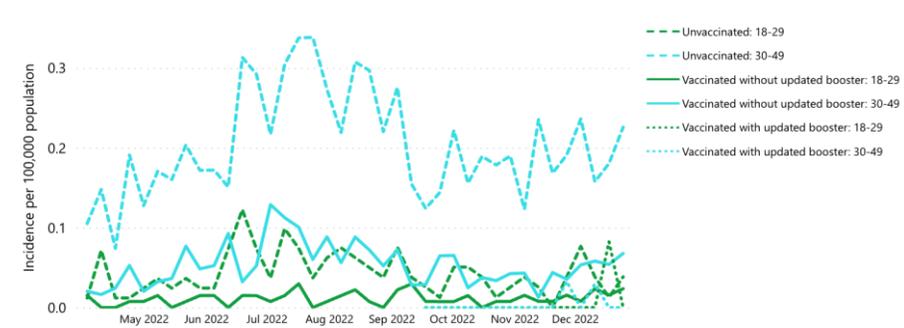
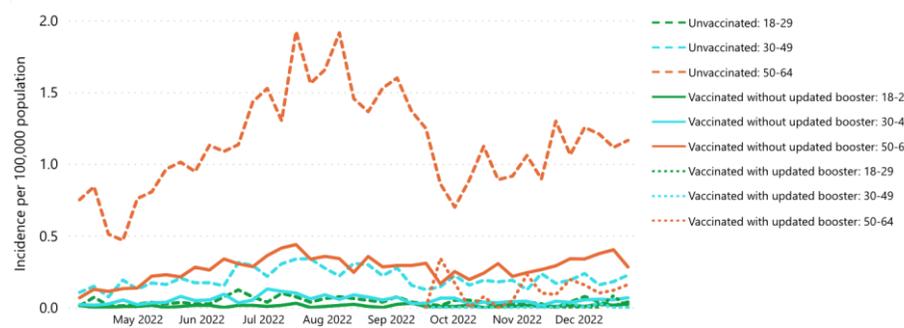
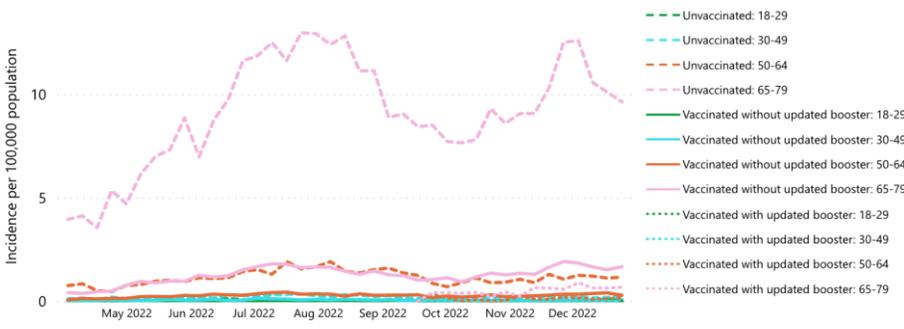
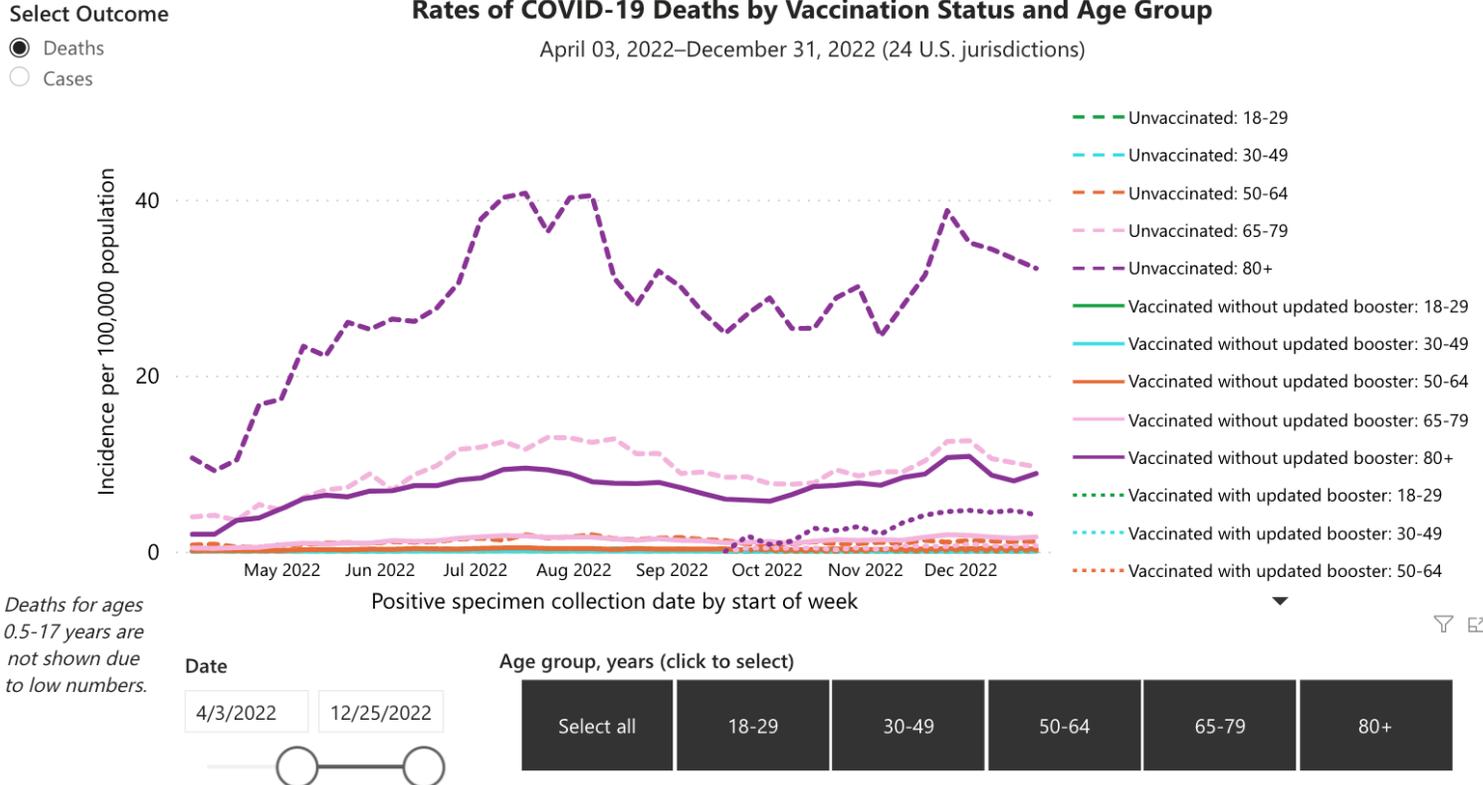
Variants Detected, by Collection Week



<https://covid.cdc.gov/covid-data-tracker/#traveler-genomic-surveillance>  
<https://github.com/gerstung-lab/SARS-CoV-2-International> (02/09/23)

# Pandemic Pubs (Feb 22<sup>nd</sup>, 2023)

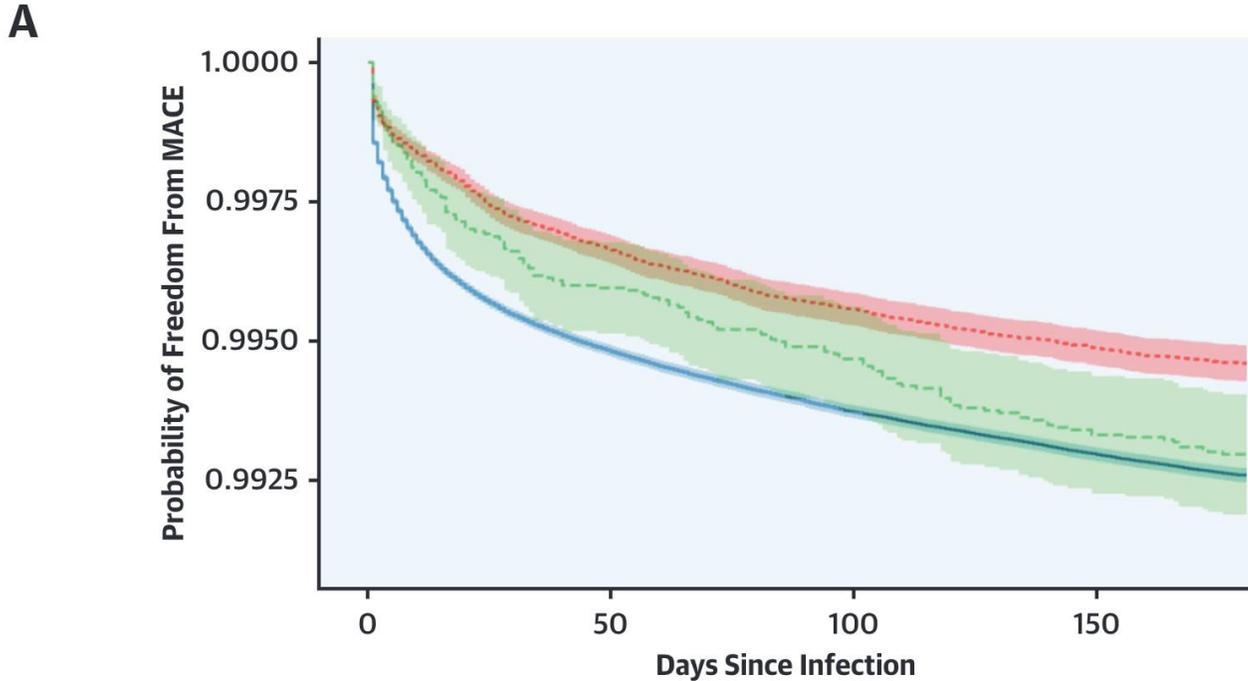
1. New data posted by the CDC show that for ages 18 and older, individuals vaccinated with a bivalent booster had 9.8x lower risk of dying from COVID-19 compared with unvaccinated people and 2.4x lower risk of dying compared to people vaccinated without updated booster



<https://covid.cdc.gov/covid-data-tracker/#rates-by-vaccine-status>

# Pandemic Pubs (Feb 22<sup>nd</sup>, 2023)

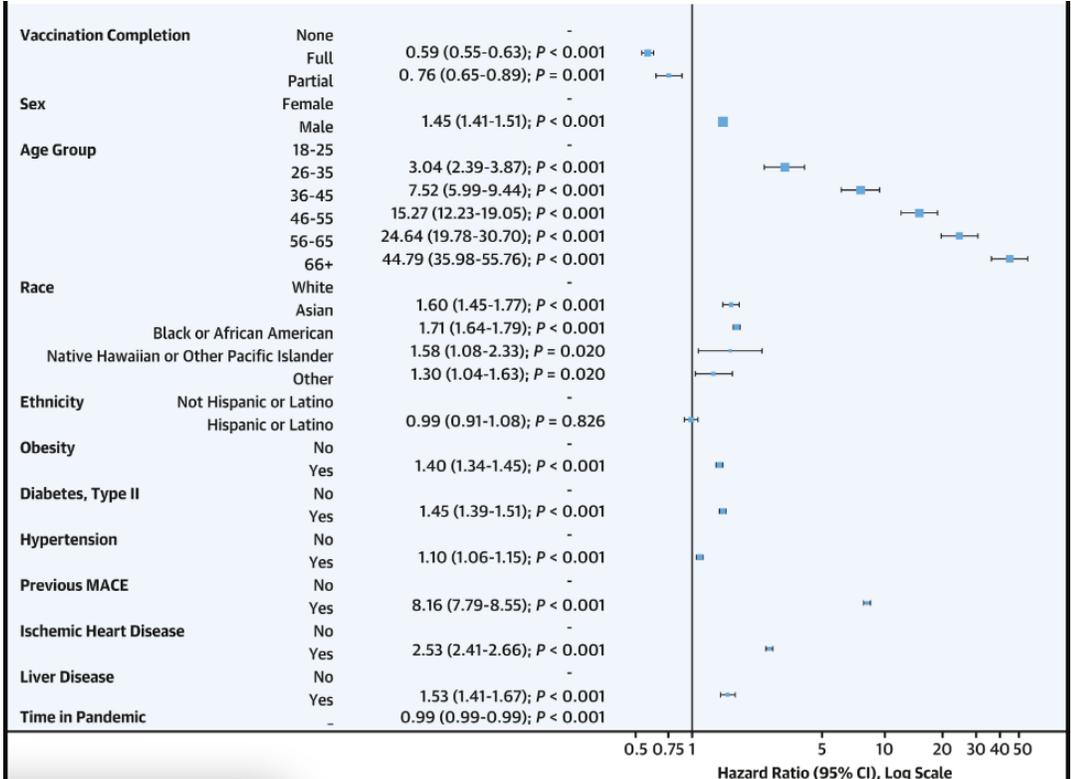
## 2. New study from Icahn school of Medicine associates vaccination with lower risk of Major Adverse Cardiac Event due to COVID



Number at risk	0	50	100	150
— No Vaccination	716,451	1,707,577	1,705,689	1,704,379
— Full Vaccination	195,136	194,484	194,272	194,136
— Partial Vaccination	22,707	22,615	22,586	22,555

<https://www.sciencedirect.com/science/article/pii/S073510972207601X>

Data was taken from the National COVID Cohort Collaborative (N3C), including 1,934,294 patients aged 18 to 90 years who were initially infected with SARS-CoV-2 between March 1, 2020, and February 1, 2022. Analysis also shows risk of COVID induced MACE significantly increased with male sex; age, notably among patients  $\geq 66$  years of age; and comorbidities, especially previous MACE.



# Influenza Update

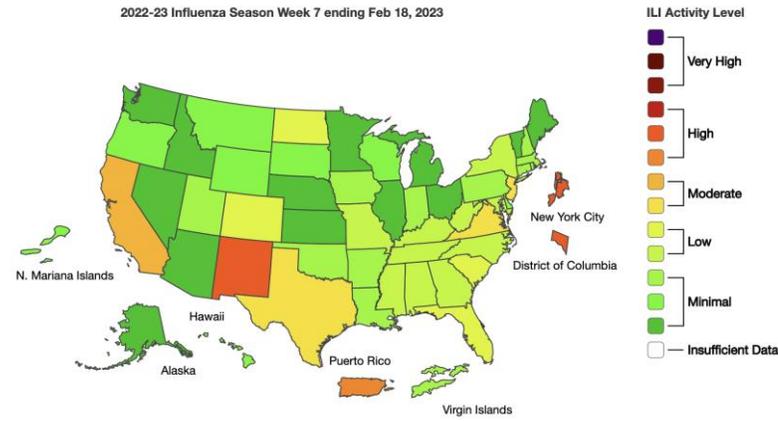
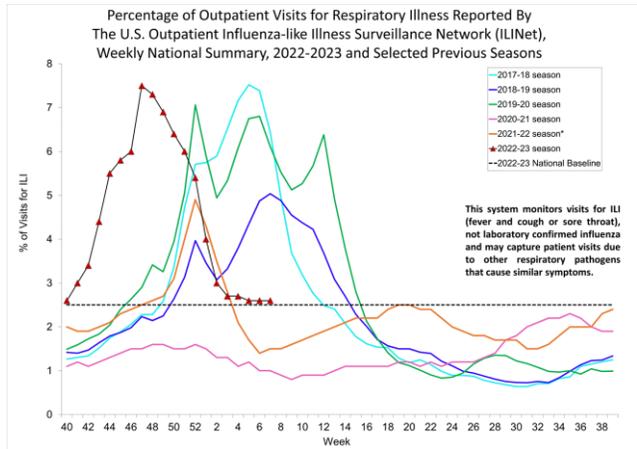
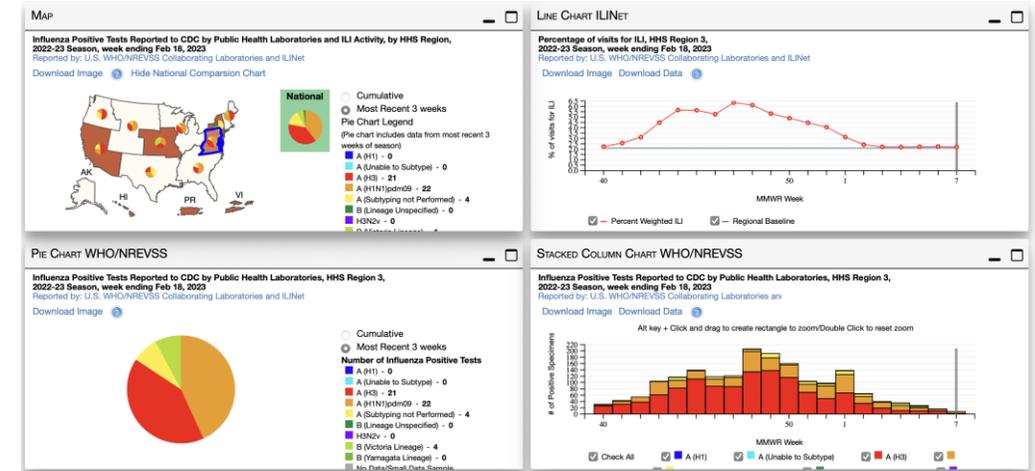
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# Current Influenza Situation – ILI Activity

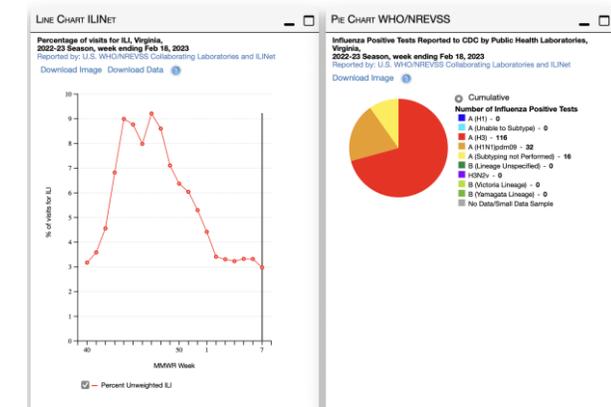
## Influenza Activity is Higher than Usual

- Virginia has shifted to “Moderate” level as most states have receded to Low and Minimal levels in the past couple weeks.
- In VA ILI Activity has declined to 3-4% which is the same as in early October at the beginning of the season
- National ILI activity has also consistently declined since a peak in late November, yet remains above threshold
- Over half of the HHS regions are now below the seasonal threshold for ILI activity

### Region 3



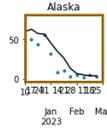
### Virginia



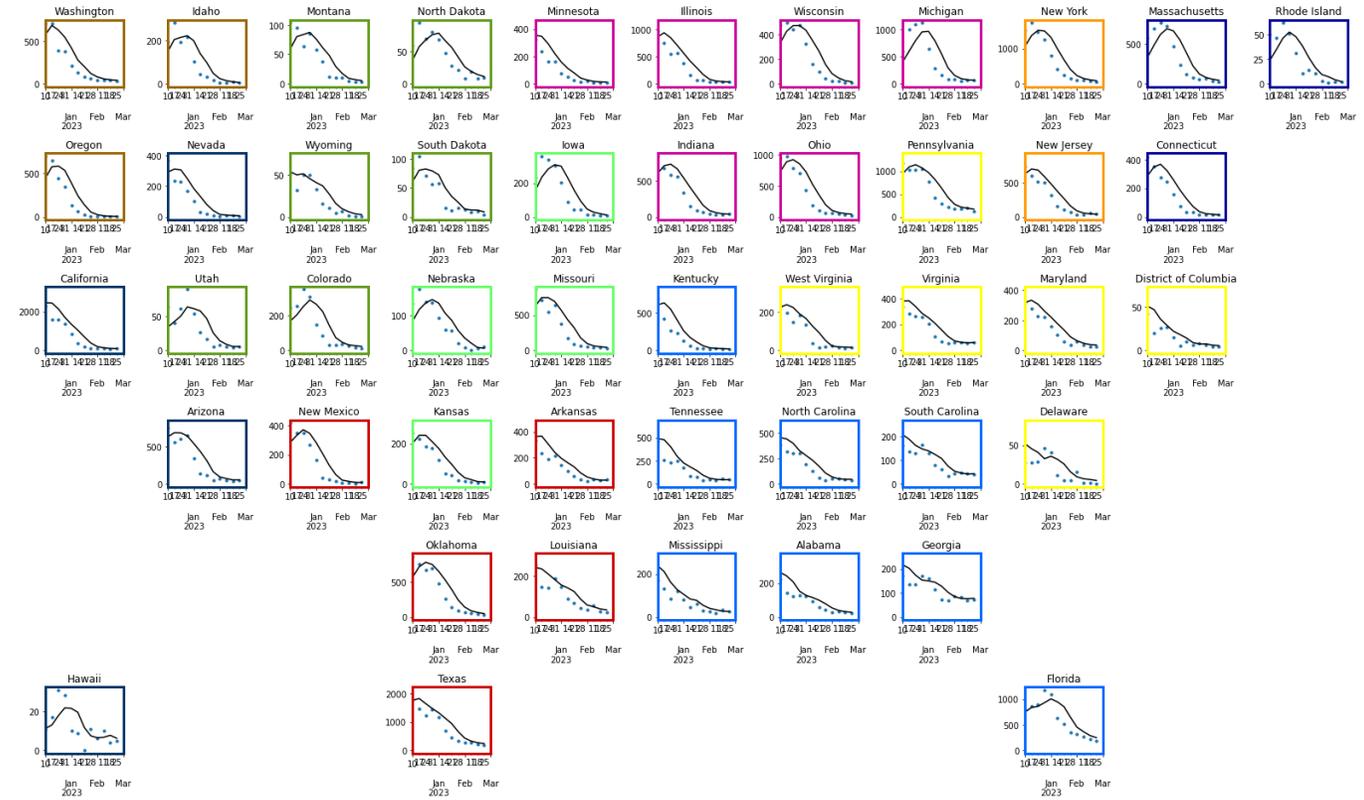
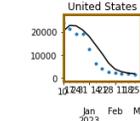
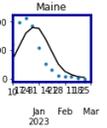
# Current Influenza Situation - Hospitalizations

## Influenza A hospitalizations continue decline

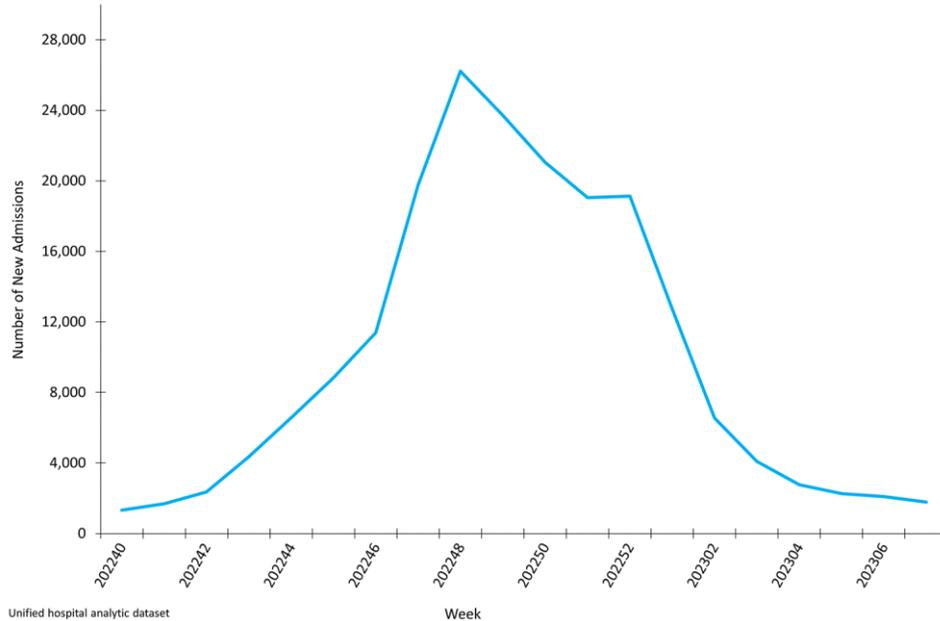
- National level of influenza hospitalizations have dropped to nearly pre-season levels
- Nearly all states have returned to levels below early December before the initial rise to to the peak



## Influenza Hospital Admissions (HHS Protect)



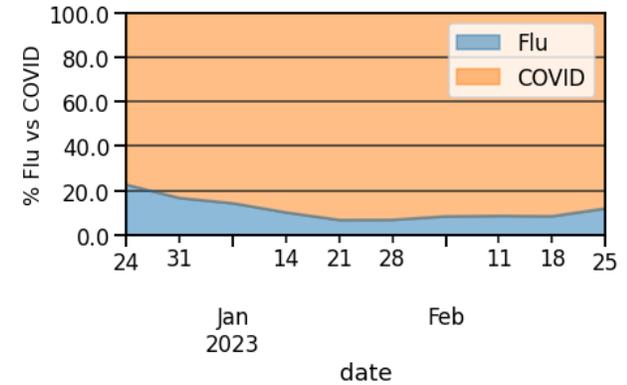
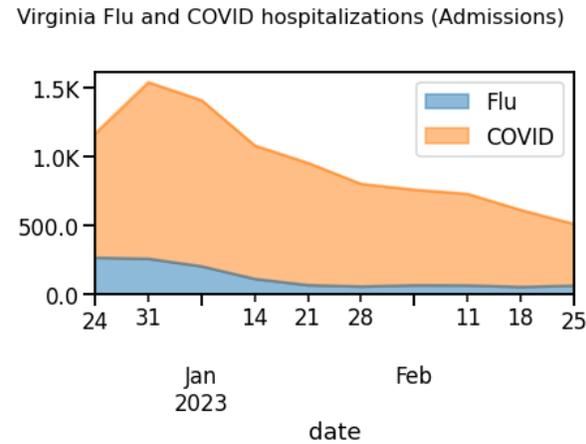
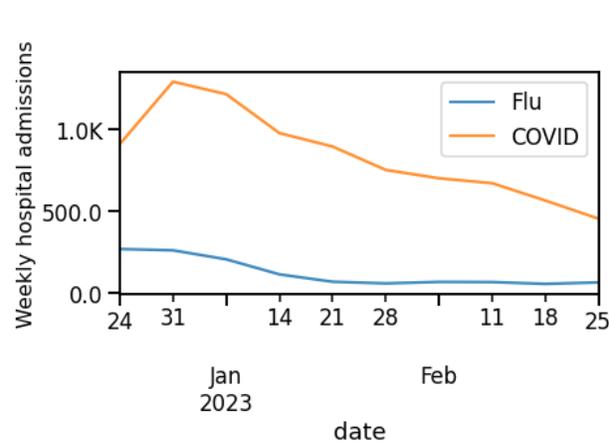
New Influenza Hospital Admissions Reported to HHS Protect, National Summary, October 2, 2022 – February 18, 2023



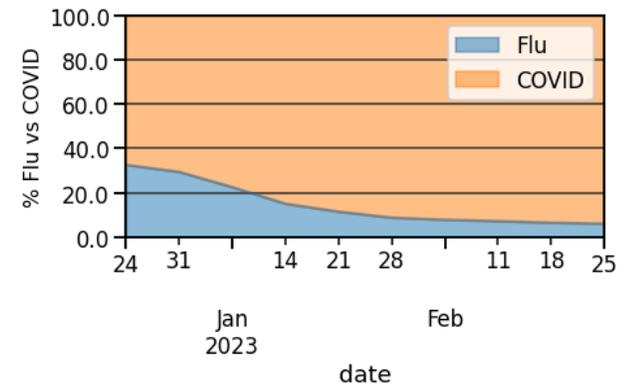
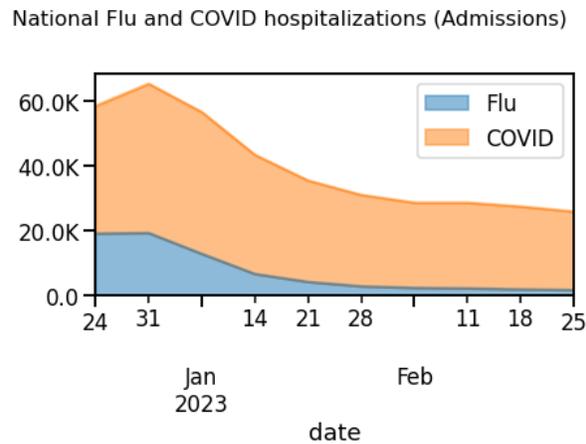
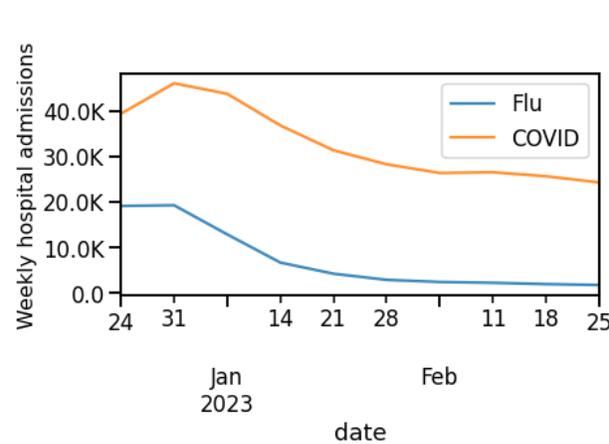
# Current Combined Hospitalizations (COVID-19 & Influenza)

## COVID-19 and Influenza Weekly Hospitalizations (HHS Protect)

Virginia



USA



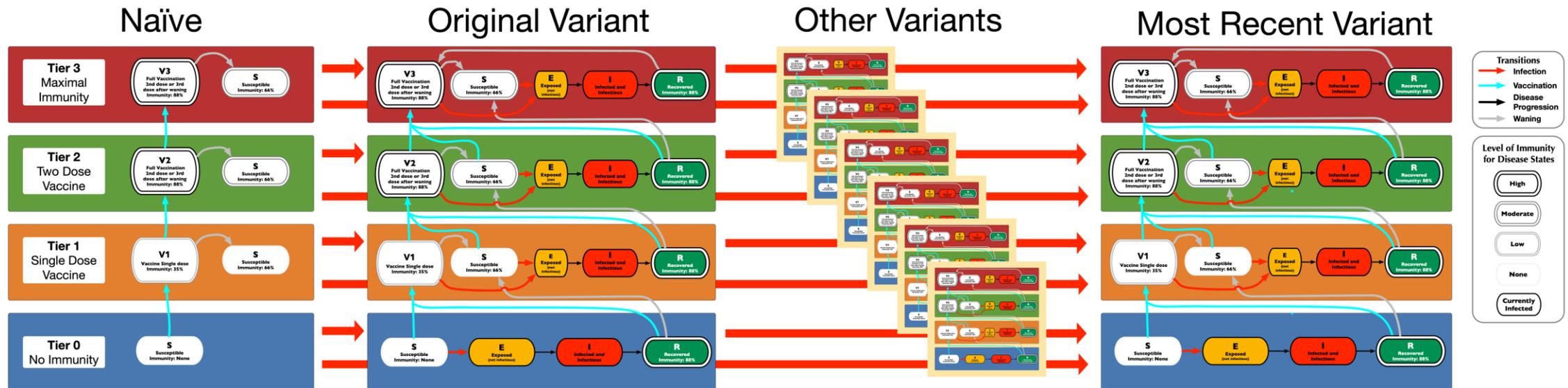
# Model Update – Adaptive Fitting

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# Model Structure Extended for more sub-variants

## Omicron sub-variants escape immunity induced by previous sub-variants

- Multiple strain support allows representation of differential protection based on immunological history (BA.1, BA.2, BA.2.12.1, BA.4/5, and future variants (VariantX) )
- Each sub-variant has differing levels of immune escape to previous sub-variants, the prevalences are based on observations for fitting purposes, and projections use estimated future prevalences
- Adaptive fitting approach continues to use simulation to generate the full distribution of immune states across the population



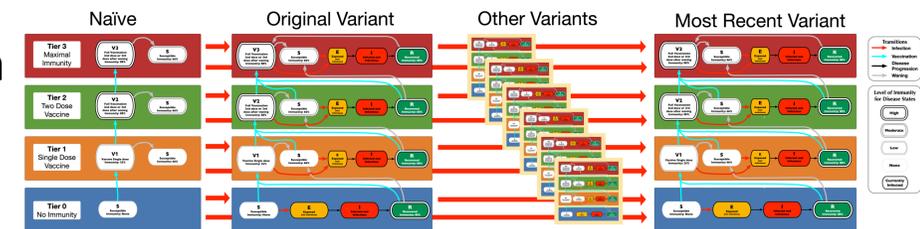
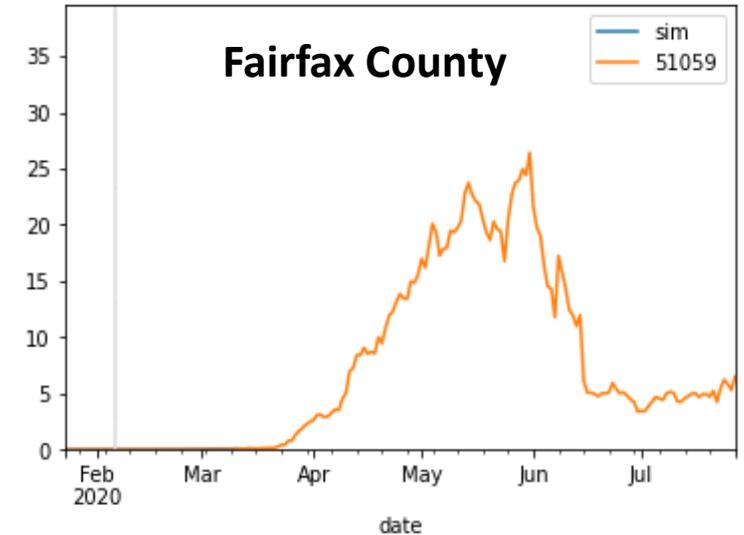
# Adaptive Fitting Approach

## Each district fit precisely, with recent trends used for future projection

- Allows history to be precisely captured, and used to guide bounds on projections

**Model:** An alternative use of the same meta-population model, PatchSim with multiple tiers of immunity

- Allows for future “what-if” Scenarios to be layered on top of calibrated model
- Allows for waning of immunity and for partial immunity against different outcomes (eg lower protection for infection than death)
- Incorporation of new variants with different levels of immune escape and transmission advantage



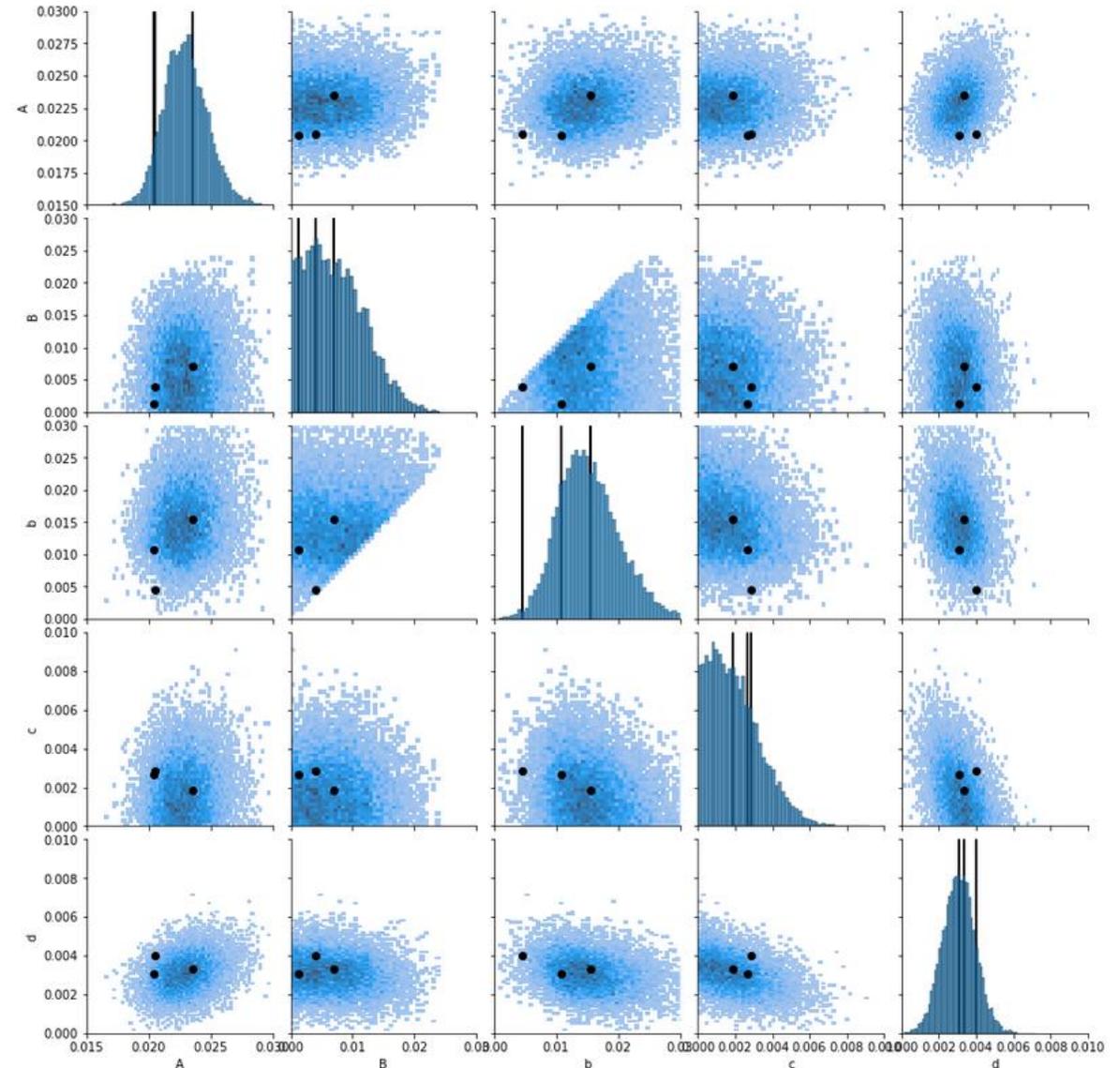
# Adaptive Fitting Approach

## Hospitalization risk for different tiers of immunity calibrated to US-wide data

- Using modeled infections from past state-level projections from Scenario Modeling Hub – using periods with reasonable confidence in ascertainment ratio (before Omicron – based on seroprevalence data, after first Omicron wave – adjusted using surveys about home test usage)
- Using all US states and bootstrapping to reduce bias and overfitting

### Selected representative risk of hospitalization per infection:

- 1 in 42-49 for unprotected (A); 1 in 142-770 for vaccinated (B);
- 1 in 64-220 for waned vaccinated (b);
- 1 in 350-540 for waned reinfection or booster(no reinfection) (c)
- 1 in 250-320 for waned vaccination/booster with reinfection (d)



# Scenarios – Transmission Conditions

- Variety of factors continue to drive transmission rates
  - Seasonal impact of weather patterns, travel and gatherings, fatigue and premature relaxation of infection control practices
- **Waning Immunity:** Omicron waning with a mean of 4 months
- **Projection Condition Ingredients:**
  - **Adaptive:** Controls remain as currently experienced into the future with NO influence from other conditions (eg seasonal, variants, etc.)
  - **Seasonal:** Controls remain the same, however, seasonal forcing or other seasonal behavior patterns
  - **New Variants (VariantX):** As of yet unidentified novel sub-variant with similar immune escape but no transmission advantage emerges 4 months after the last significant sub-variant and grows at a similar rate

# Projection Scenarios – Combined Conditions

Name	Txm	Variant	Booster	Description
Adaptive-VariantX	C	X	Current	Like Adaptive, with emergence of a Variant like XBB.1.5 that tracks its prevalence
Adaptive-VariantX-IncreasePerm	Increase	X	Current	Like Adaptive-VariantX but with an increase of 30% over the course of 4 weeks, that remains constant thereafter
Adaptive-VariantX-IncreaseTemp	Increase	X	Current	Like Adaptive-VariantX but with an increase of 30% over the course of 4 weeks and then recedes over the course of 4 weeks

**Transmission:**

C = Current levels persist into the future

Increase = Transmission rates increase a total of 30% over 4 weeks representing a delayed seasonally or variant driven bump, this in effect returns transmission rates to similar levels as last summer

**Variant:**

SQ = Status quo prevalences remain the same (e.g. no significant major driving of transmission anticipated)

X = Novel sub-variant scenario, new variant emerges reaches dominance in near term, 30% immune escape

**Booster:**

Current = Current pace relative to 3<sup>rd</sup> dose rollout is maintained in the future

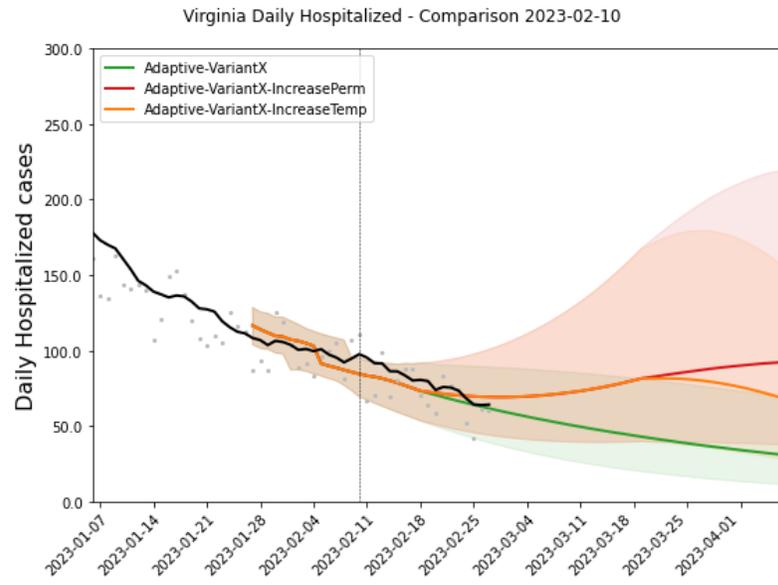
# Model Results

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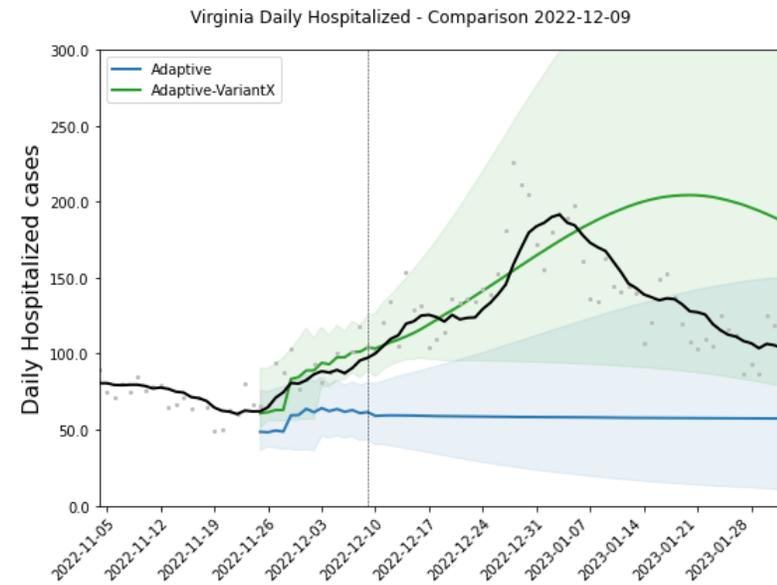
# Previous projections comparison - Hospitalizations

- Previous projections have tracked observed hospitalizations well
- Projection from 2 weeks ago had enough growth in hospitalizations to track well
- Projection from 4 weeks ago had declines which delayed the growth
- Projection from early July anticipated a Fall-Winter rise that has tracked well

Previous round – 2 weeks ago



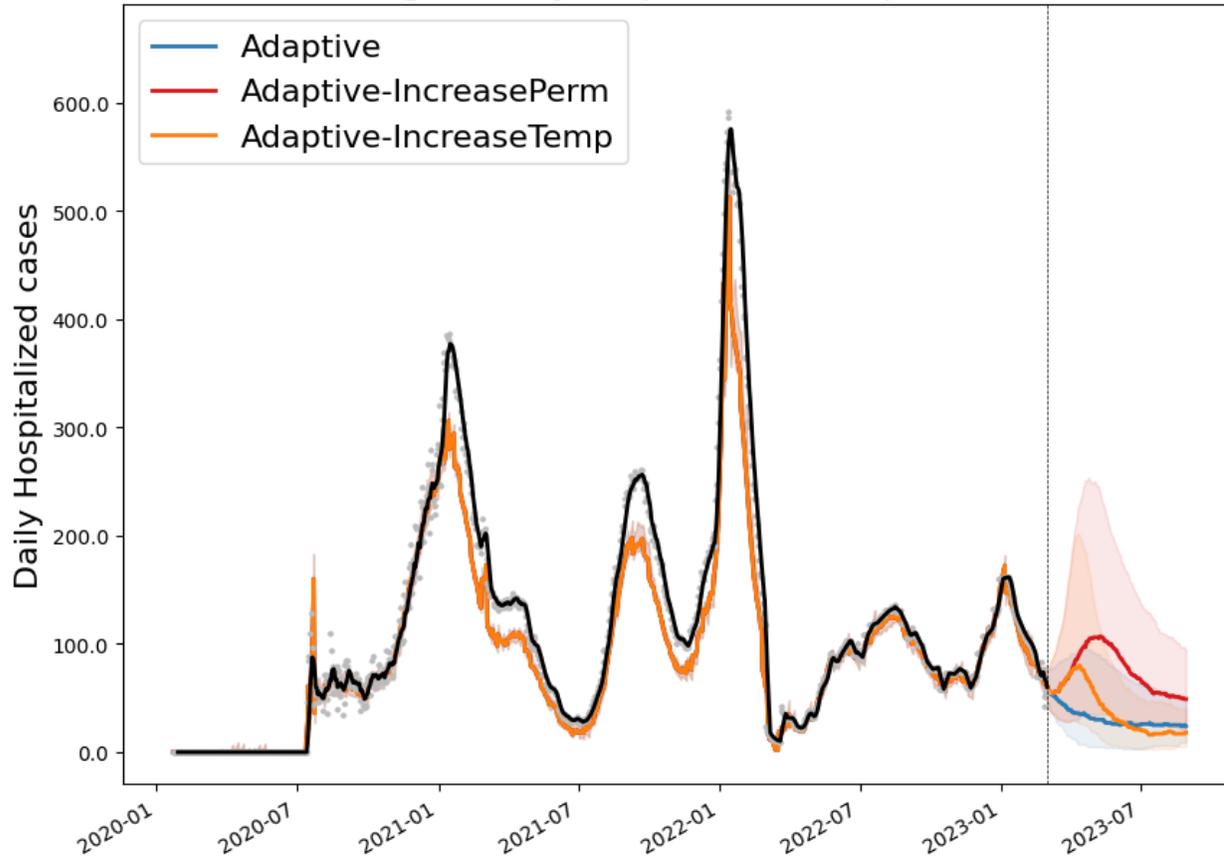
Previous round – mid December



# Outcome Projections

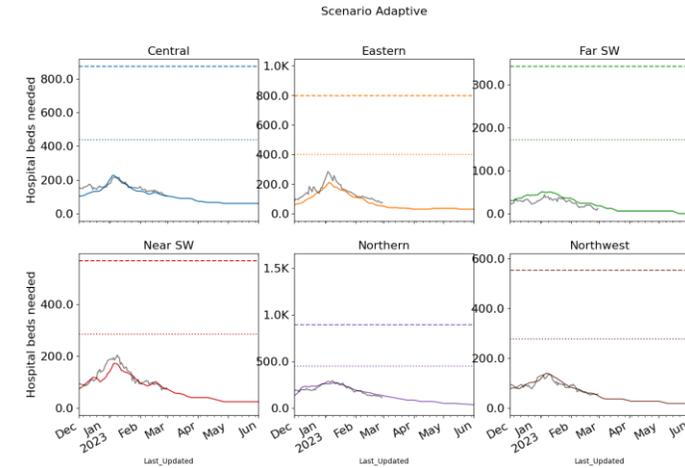
## Daily Hospitalized

Virginia Daily Hospitalized - Comparison

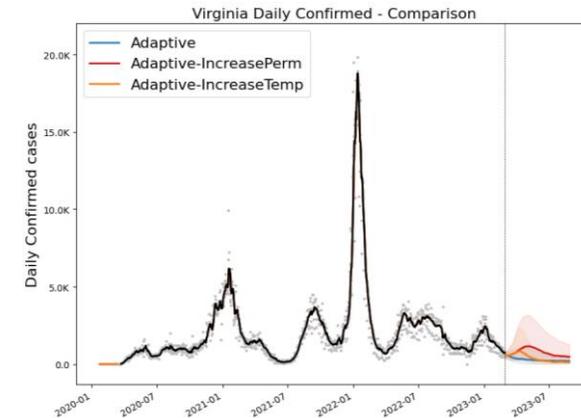


\* without surveillance correction VariantBA2 peaked over 10K in July

## Estimated Hospital Occupancy

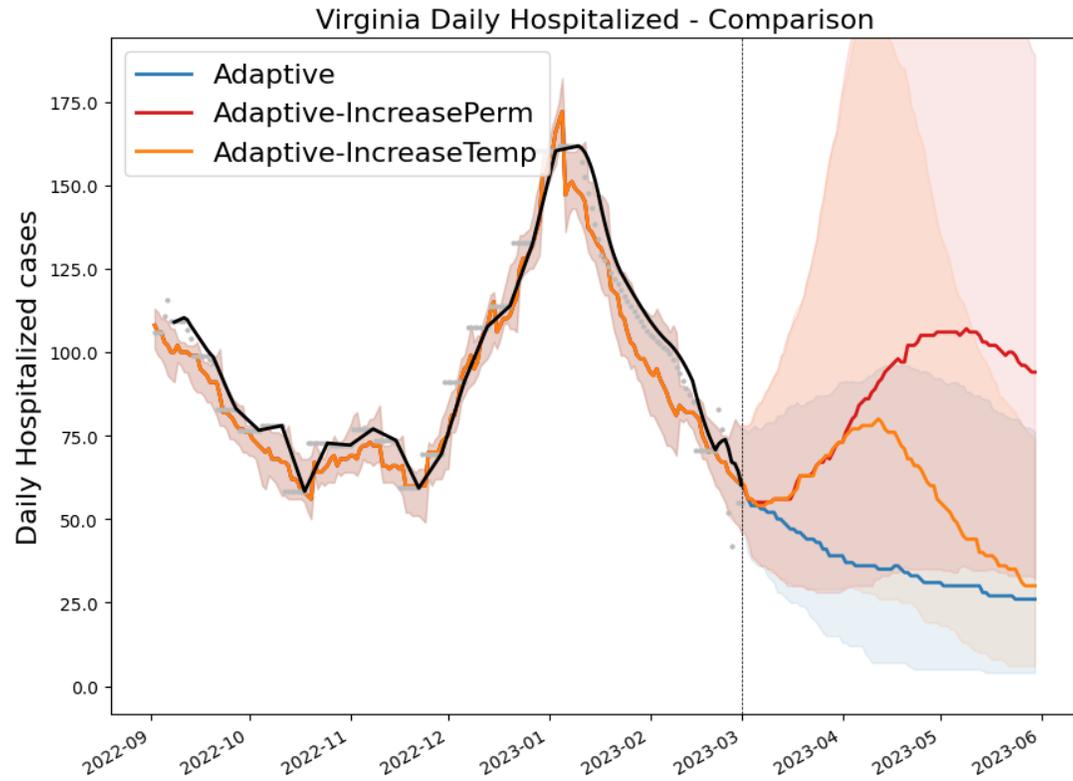


## Confirmed cases

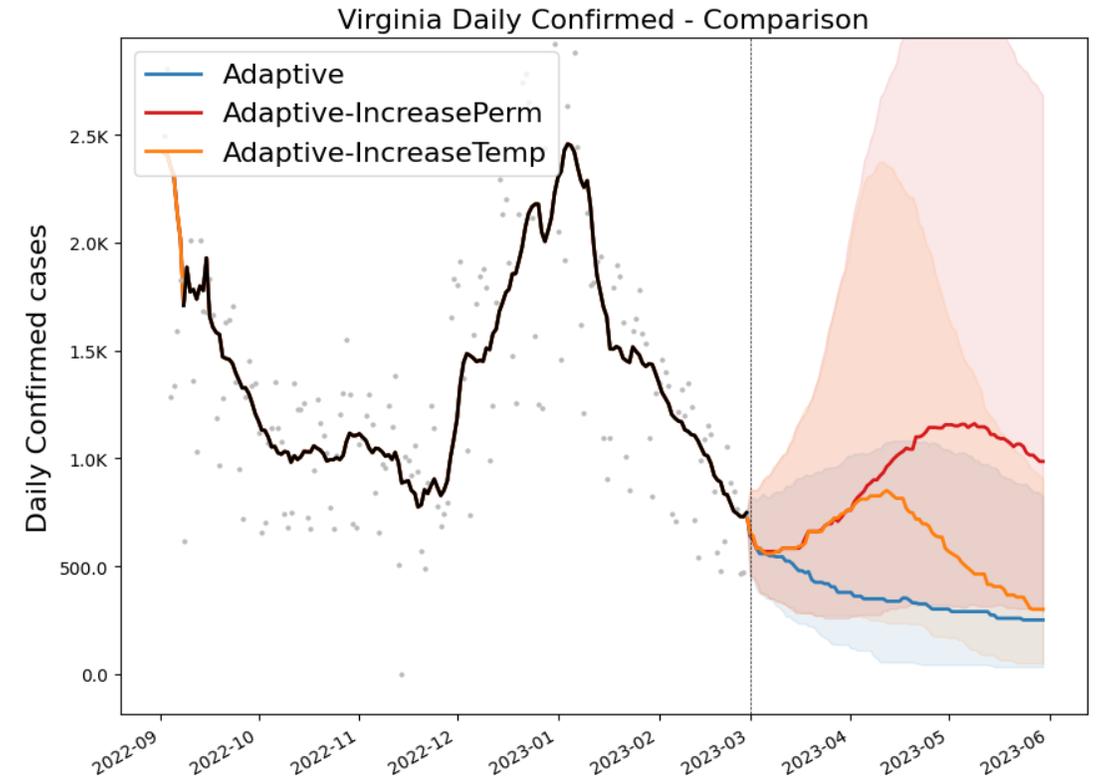


# Outcome Projections – Closer Look

## Daily Hospitalized

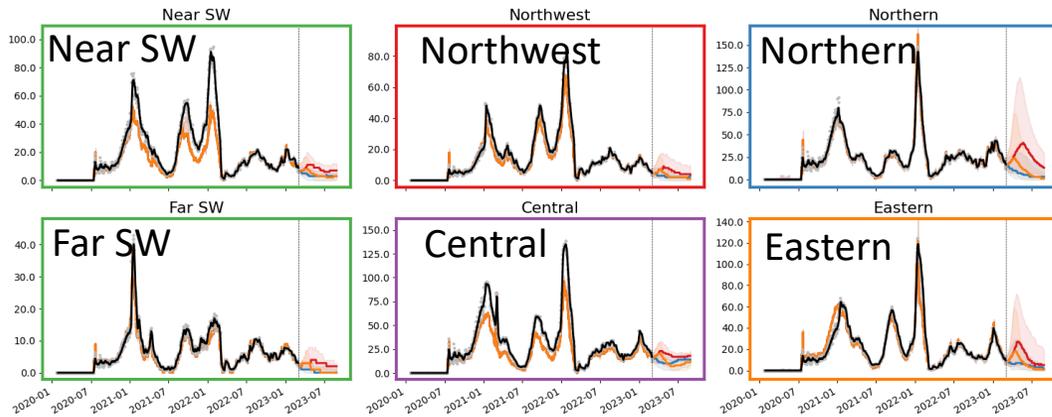


## Confirmed cases

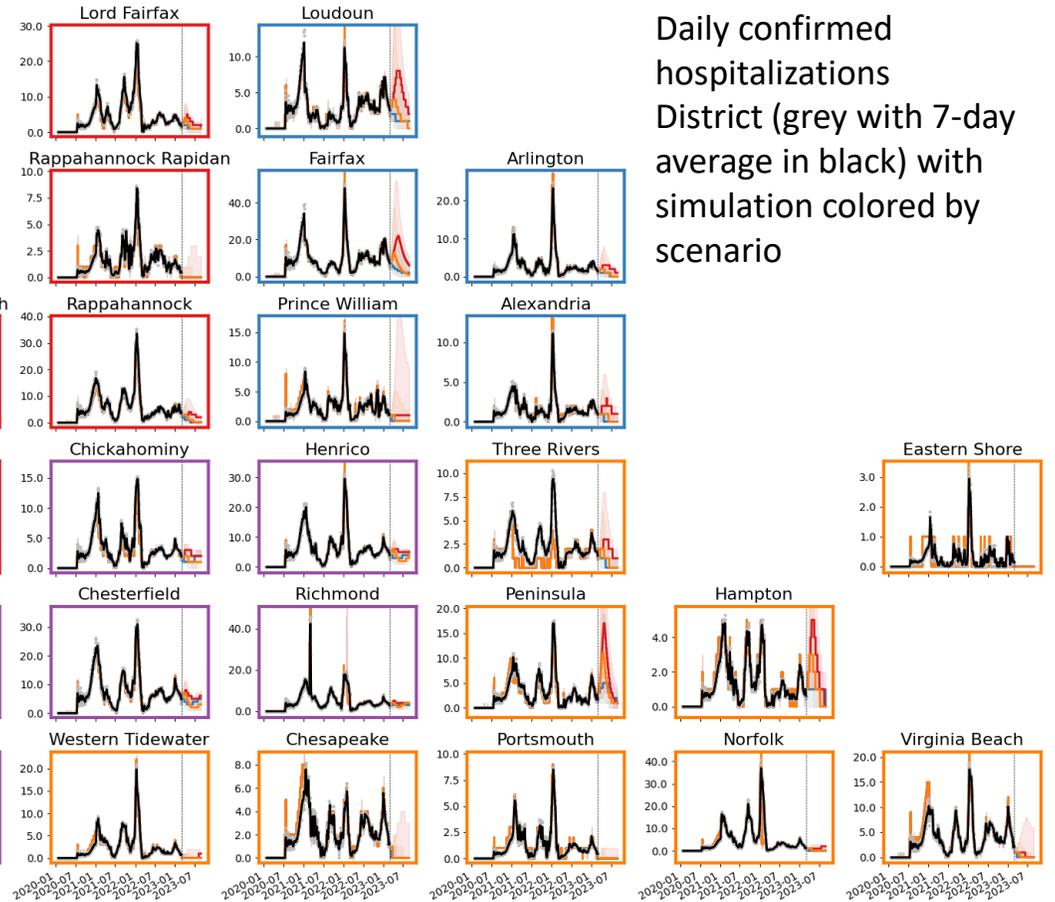


# Detailed Projections: Hosps for All Scenarios

## Projections by Region



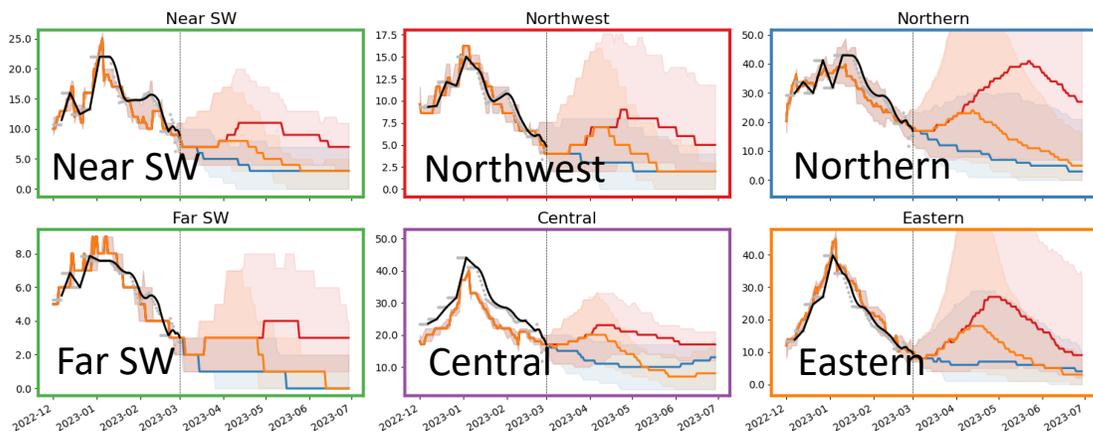
## Projections by District



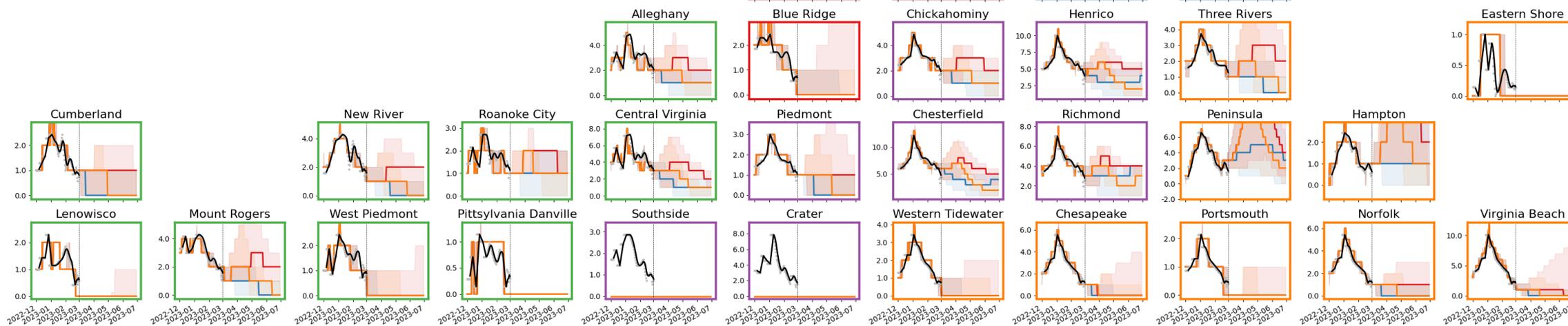
Daily confirmed hospitalizations District (grey with 7-day average in black) with simulation colored by scenario

# Detailed Projections: Hosps for All Scenarios - Closer Look

## Projections by Region



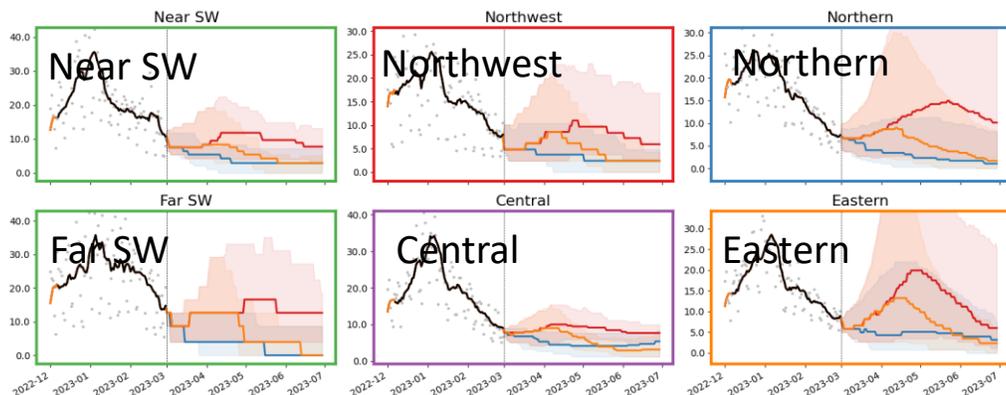
## Projections by District



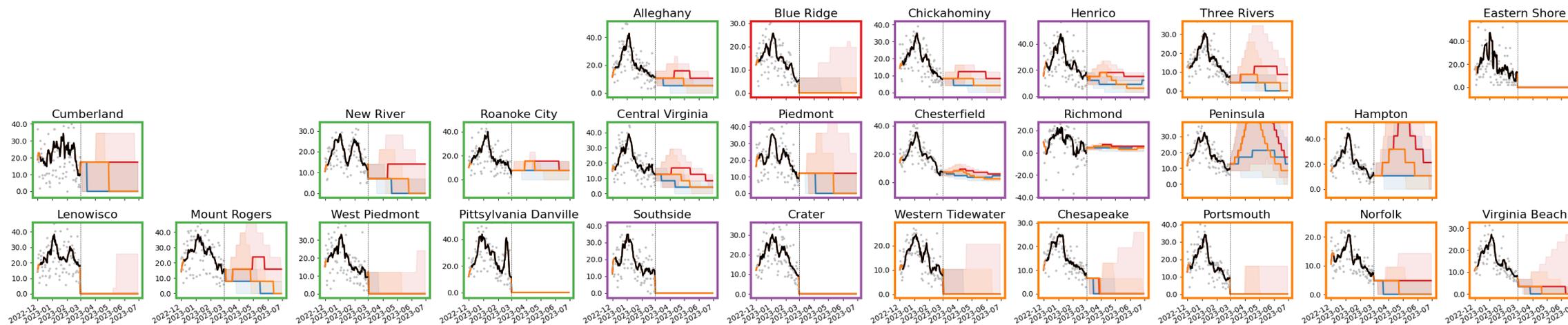
Daily confirmed hospitalizations District (grey with 7-day average in black) with simulation colored by scenario

# Detailed Projections: Cases for All Scenarios - Closer Look

## Projections by Region



## Projections by District



Daily confirmed cases by rate (per 100K) District (grey with 7-day average in black) with simulation colored by scenario

# National Modeling Hub Updates

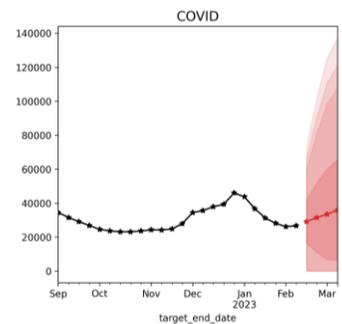
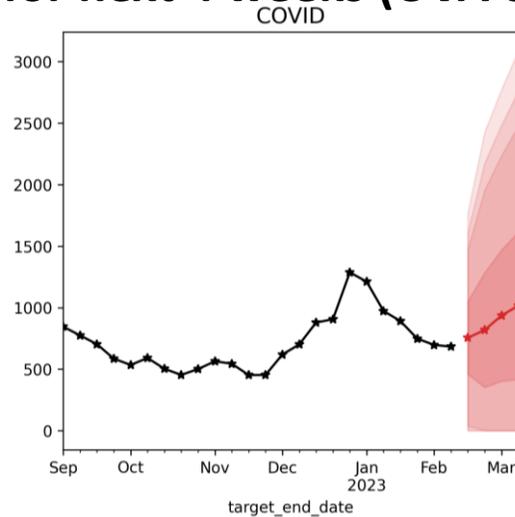
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# Current COVID-19 Hospitalization Forecast

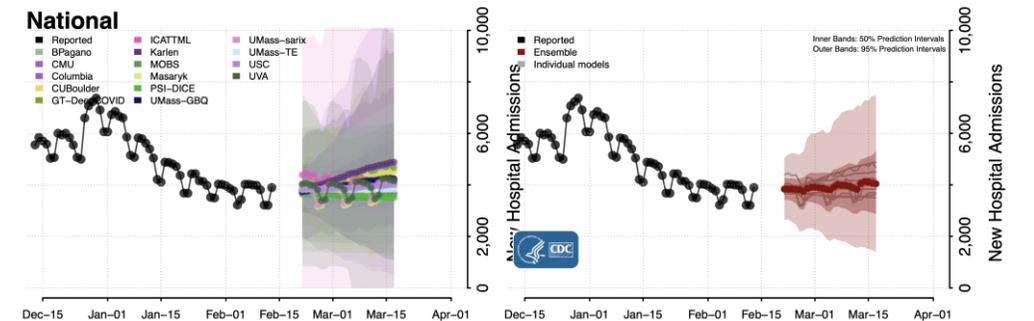
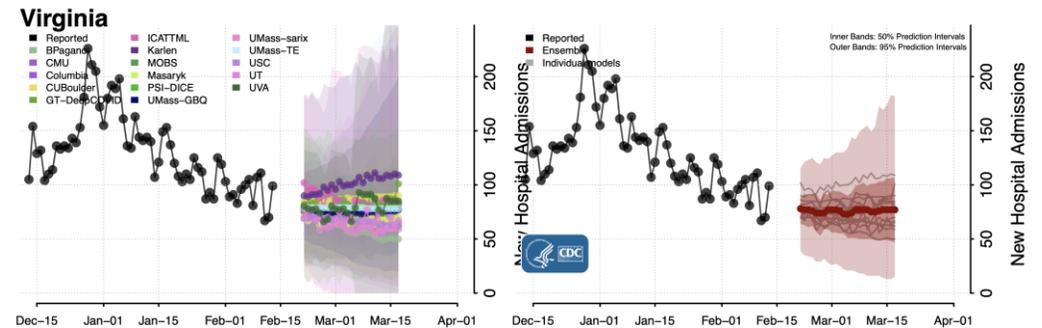
## Statistical models for submitting to CDC FluSight forecasting challenge

- Uses a variety of statistical and ML approaches to forecast weekly hospital admissions for the next 4 weeks for all states in the US

### Hospital Admissions for COVID-19 and Forecast for next 4 weeks (UVA ensemble)



### Hospital Admissions for COVID-19 and Forecast for next 4 weeks (CDC COVID Ensemble)

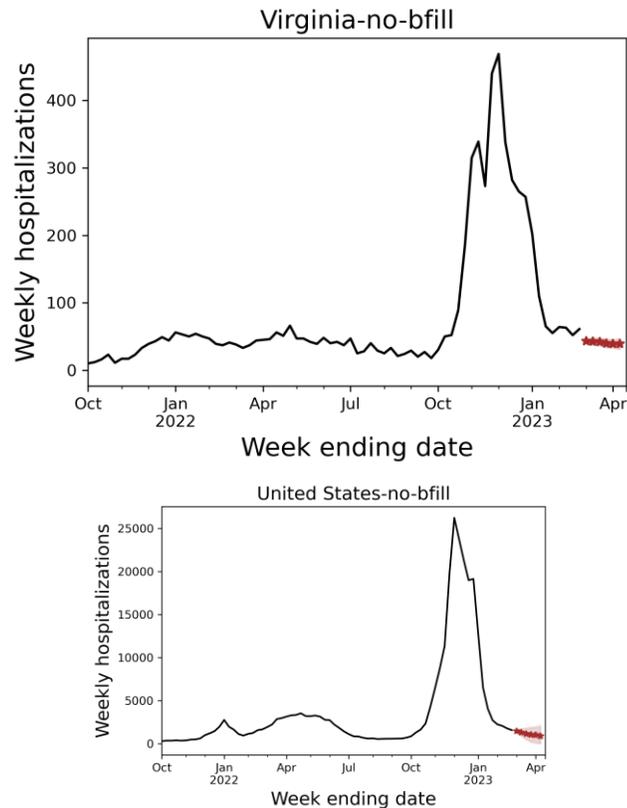


# Current Influenza Hospitalization Forecast

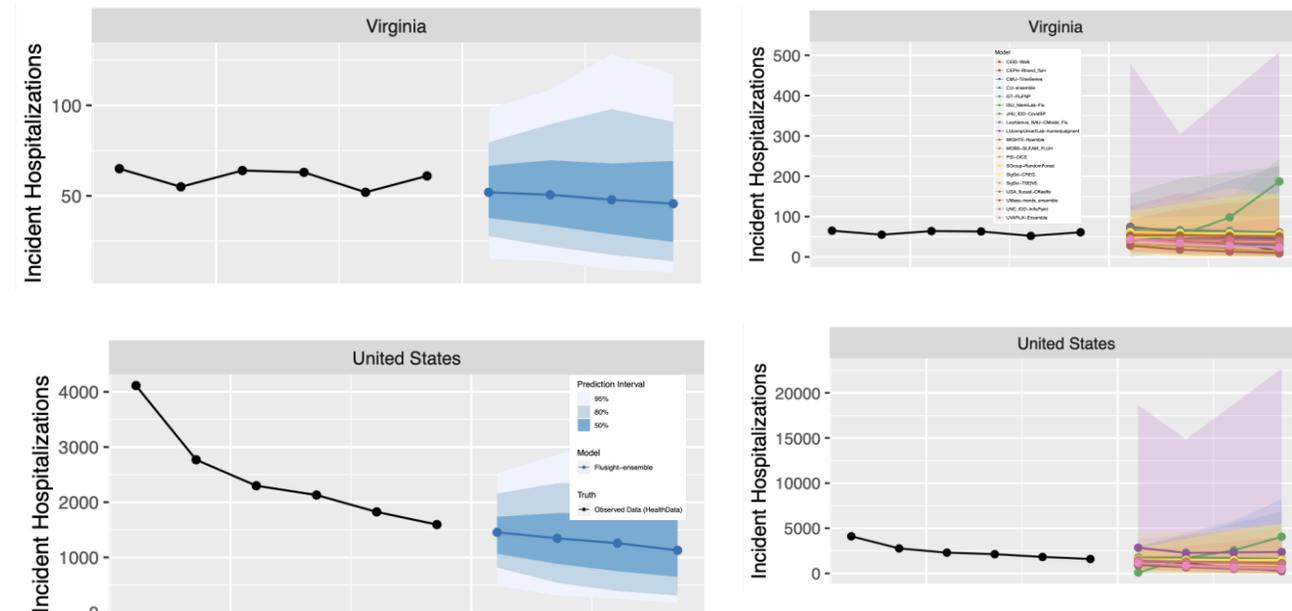
## Statistical models for submitting to CDC FluSight forecasting challenge

- Similar to COVID-19 case forecasts, uses a variety of statistical and ML approaches to forecast weekly hospital admissions for the next 4 weeks for all states in the US

### Hospital Admissions for Influenza and Forecast for next 4 weeks (UVA ensemble)



### Hospital Admissions for Influenza and Forecast for next 4 weeks (CDC FluSight Ensemble)



# Combined ILI and COVID-19 Hospitalizations

Ensemble methodology that combines the Adaptive with machine learning and statistical models such as:

- Autoregressive (AR, ARIMA), Neural networks (LSTM), Kalman filtering (EnKF), G-model (phase), Holt-Winters

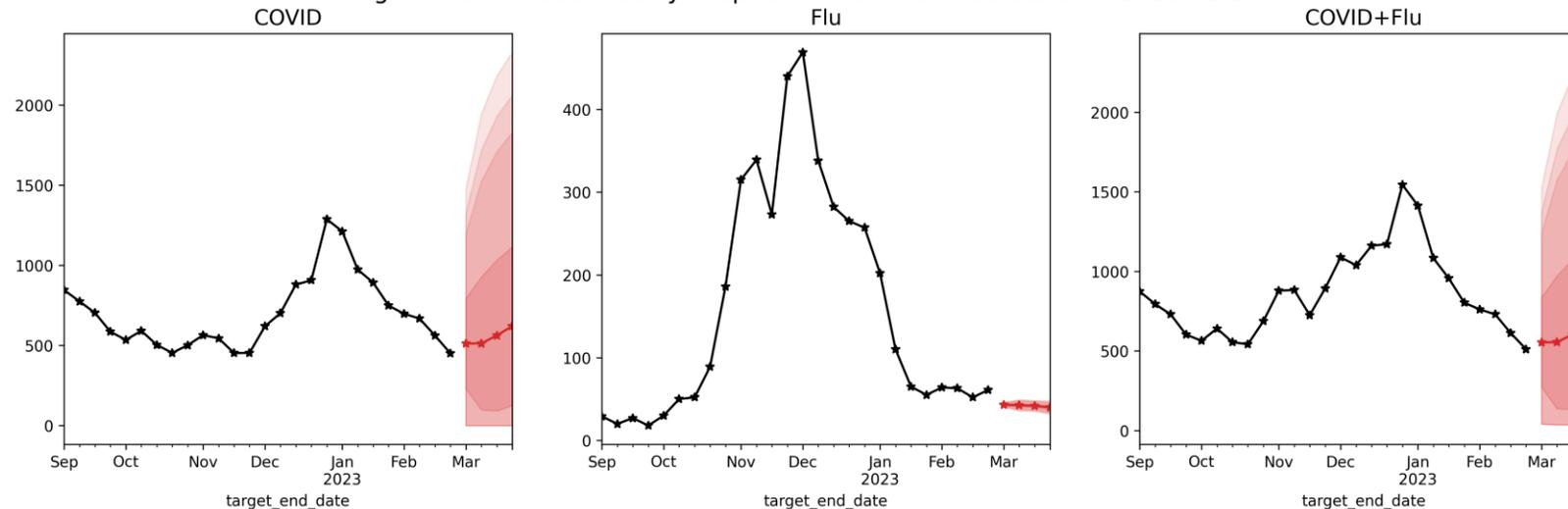
Weekly forecasts of hospitalizations done at state level.

Models chosen because of their track record in disease forecasting and to increase diversity and robustness.

Both are regularly submitted to CDC Forecast Hubs

## Weekly Hospitalizations Short-term COVID-19 and Influenza Forecasts

Virginia - UVA model weekly hospital admissions forecasts as of 2023-02-27





# Scenario Modeling Hub – COVID-19 (Round 16)

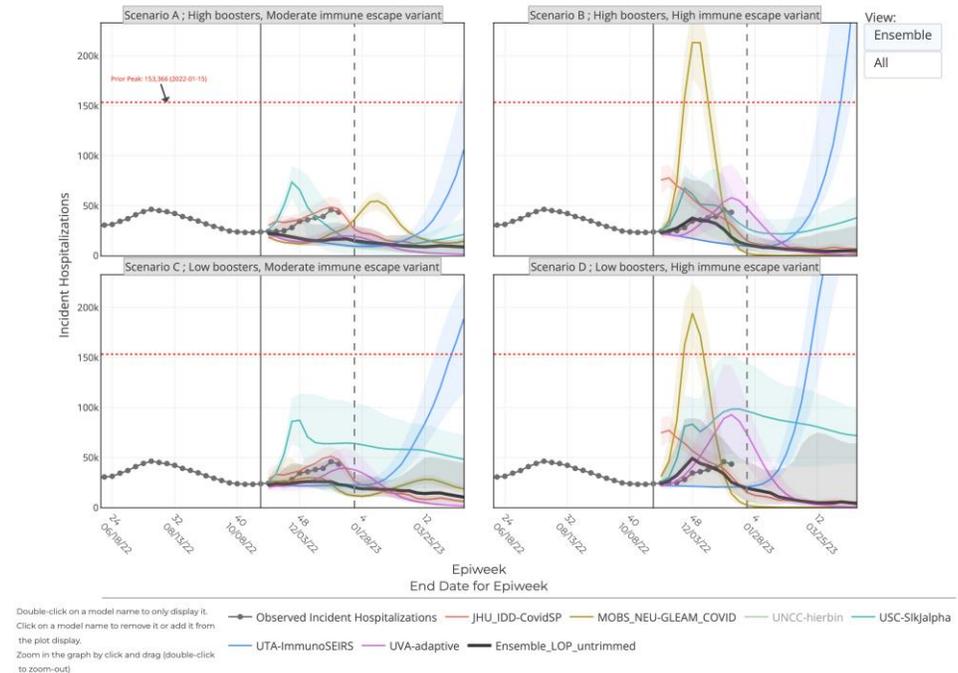
Collaboration of multiple academic teams to provide national and state-by-state level projections for 4 aligned scenarios

- Round 16 results published
- Moderate escape scenarios tracking best

<https://covid19scenariomodelinghub.org/viz.html>

	"Level 5" Variants	"Level 6/7" Variants
Accelerating uptake levels of reformulated boosters	<p><b>Scenario A</b></p> <p>"Level 5" Variants</p> <ul style="list-style-type: none"> <li>- Variants have a 25% immune escape from BA.5.2</li> <li>- Seeding based on combined observed prevalence of Level 5 variants at the start of the projection period</li> <li>- No change in severity given symptomatic infection</li> </ul> <p>Accelerating uptake levels of reformulated boosters, with coverage plateauing at 90% of flu vaccination levels by February 1st, 2023</p> <ul style="list-style-type: none"> <li>- Teams are free to use available data and information from current and previous rollouts as they see fit to define rates</li> <li>- Teams should assume increasing uptake through October and November as necessary to reach the projected February 1st, 2022 plateau</li> </ul>	<p><b>Scenario B</b></p> <p>"Level 6/7" Variants</p> <ul style="list-style-type: none"> <li>- Variants have a 50% immune escape from BA.5.2</li> <li>- Seeding based on combined observed prevalence of Level 6 and 7 variants at the start of the projection period</li> <li>- No change in severity given symptomatic infection</li> </ul> <p>Accelerating uptake levels of reformulated boosters, with coverage plateauing at 90% of flu vaccination levels by February 1st, 2023</p> <ul style="list-style-type: none"> <li>- Teams are free to use available data and information from current and previous rollouts as they see fit to define rates</li> <li>- Teams should assume increasing uptake through October and November as necessary to reach the projected February 1st, 2022 plateau</li> </ul>
Current uptake levels of reformulated boosters	<p><b>Scenario C</b></p> <p>"Level 5" Variants</p> <ul style="list-style-type: none"> <li>- Variants have a 25% immune escape from BA.5.2</li> <li>- Seeding based on combined observed prevalence of Level 5 variants at the start of the projection period</li> <li>- No change in severity given symptomatic infection</li> </ul> <p>Current uptake levels of reformulated boosters, with coverage plateauing at booster 1 levels by the end of the simulation</p> <ul style="list-style-type: none"> <li>- Teams are free to use available data and information from current and previous rollouts as they see fit to define rates</li> <li>- Based on current rates, plateau date is flexible as long as it occurs before the end of the simulation (Teams can adjust rates up if needed to achieve adequate coverage by target date)</li> </ul>	<p><b>Scenario D</b></p> <p>"Level 6/7" Variants</p> <ul style="list-style-type: none"> <li>- Variants have a 50% immune escape from BA.5.2</li> <li>- Seeding based on combined observed prevalence of Level 6 and 7 variants at the start of the projection period</li> <li>- No change in severity given symptomatic infection</li> </ul> <p>Current uptake levels of reformulated boosters, with coverage plateauing at booster 1 levels by the end of the simulation</p> <ul style="list-style-type: none"> <li>- Teams are free to use available data and information from current and previous rollouts as they see fit to define rates</li> <li>- Based on current rates, plateau date is flexible as long as it occurs before the end of the simulation (Teams can adjust rates up if needed to achieve adequate coverage by target date)</li> </ul>

Projected Incident Hospitalizations by Epidemiological Week and by Scenario for Round 16 - US  
(- Projection Epiweek; -- Current Week)



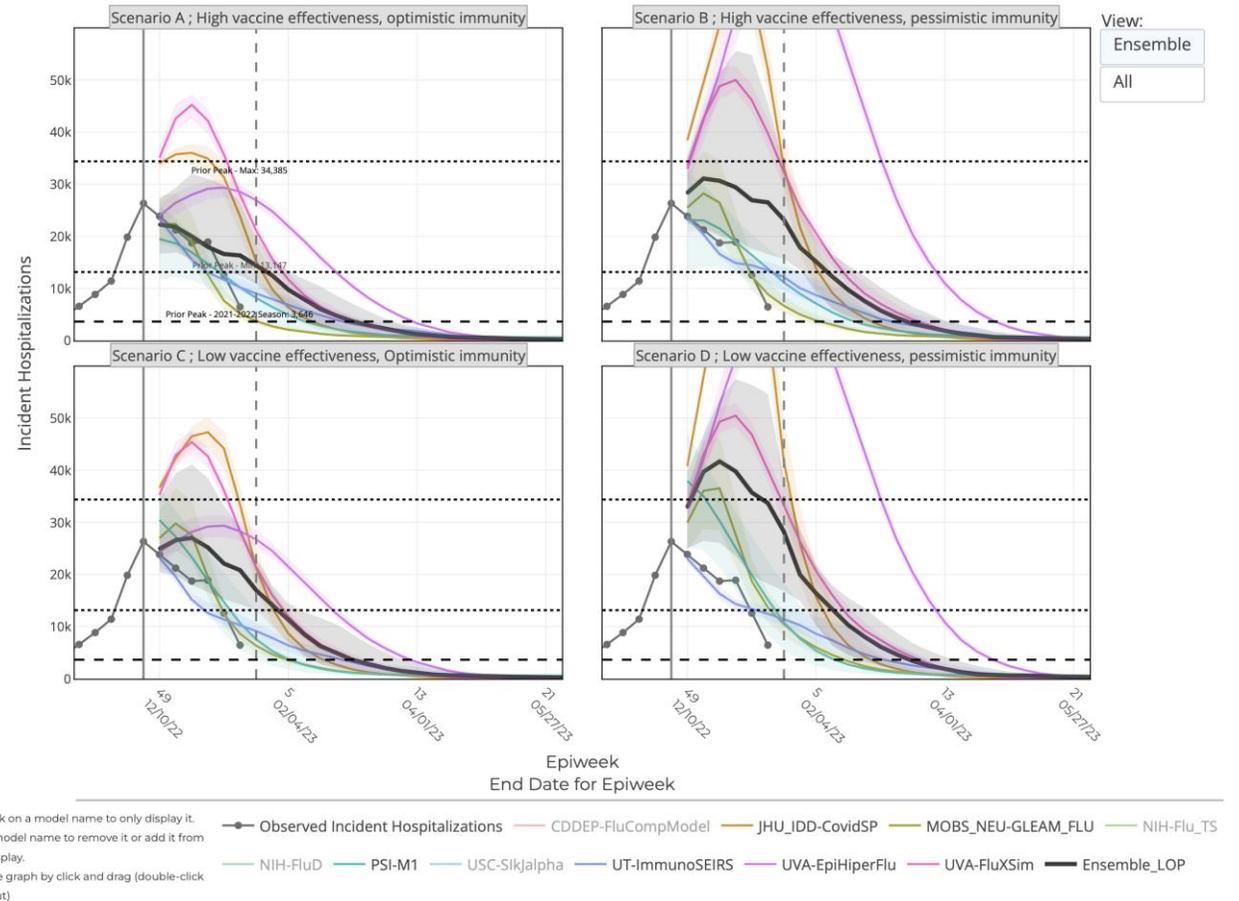
# Scenario Modeling Hub – Influenza (Round 3)

Collaboration of multiple academic teams to provide national and state-by-state level projections for 4 aligned scenarios

- All rounds so far have explored the combination of a prior immunity axis and a vaccine effectiveness axis
- Round 2 and 3 are identical in design (Round 3 cutoff December 3<sup>rd</sup>)

<https://fluscenariomodelinghub.org/viz.html>

Projected Incident Hospitalizations by Epidemiological Week and by Scenario for Round 3 - US  
(- Projection Epiweek; -- Current Week)



	Optimistic flu prior immunity	Pessimistic flu prior immunity
High Vaccine Effectiveness	<p><b>Scenario A</b></p> <p>Optimistic flu prior immunity</p> <ul style="list-style-type: none"> <li>- No impact of missed flu seasons due to the COVID-19 pandemic on prior immunity.*</li> <li>- Same amount of prior immunity as in a typical, pre-COVID19 pandemic prior season.</li> </ul> <p>High Vaccine Effectiveness</p> <ul style="list-style-type: none"> <li>- VE = 50% against medically attended influenza illnesses and hospitalizations (comparable to 2015-16 season).</li> </ul>	<p><b>Scenario B</b></p> <p>Pessimistic flu prior immunity</p> <ul style="list-style-type: none"> <li>- Substantial impact of missed flu seasons due to the COVID-19 pandemic on prior immunity.*</li> <li>- 50% lower immunity than a typical, pre-COVID19 pandemic season.</li> </ul> <p>High Vaccine Effectiveness</p> <ul style="list-style-type: none"> <li>- VE = 50% against medically attended influenza illnesses and hospitalizations (comparable to 2015-16 season).</li> </ul>
Low Vaccine Effectiveness	<p><b>Scenario C</b></p> <p>Optimistic flu prior immunity</p> <ul style="list-style-type: none"> <li>- No impact of missed flu seasons due to the COVID-19 pandemic on prior immunity.*</li> <li>- Same amount of prior immunity as in a typical, pre-COVID19 pandemic prior season.</li> </ul> <p>Low Vaccine Effectiveness</p> <ul style="list-style-type: none"> <li>- VE = 30% against medically attended influenza illnesses and hospitalizations (comparable to 2018-19 season).</li> </ul>	<p><b>Scenario D</b></p> <p>Pessimistic flu prior immunity</p> <ul style="list-style-type: none"> <li>- Substantial impact of missed flu seasons due to the COVID-19 pandemic on prior immunity.*</li> <li>- 50% lower immunity than a typical, pre-COVID19 pandemic season.</li> </ul> <p>Low Vaccination Protection</p> <ul style="list-style-type: none"> <li>- VE = 30% against medically attended influenza illnesses and hospitalizations (comparable to 2018-19 season).</li> </ul>

# Key Takeaways

Projecting future cases precisely is impossible and unnecessary.

Even without perfect projections, we can confidently draw conclusions:

- Case rates and hospitalizations from COVID-19 continue to decline
- Case rates and hospitalizations from Influenza are very low
- **Model Updates**
  - Model updated to fit hospital admissions as opposed to cases
    - New technique used to determine hospitalization to infection rate
  - Projection model updated this week, two non-specific scenarios related to increases in transmissibility.
  - Boosted transmissibility can generate new surge in activity which does not exceed levels from Summer of 2022

# Questions?

## Biocomplexity COVID-19 Response Team

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